

**Assignment -2**  
**PYTHON PROGRAM**

Assignment Date	25 SEP 2022
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Student Roll Number	610819106036
Maximum Marks	2 Marks

**Question-1:**

Download the dataset: [Dataset](#)

**Solution:**

**DATA PROCESSING**

**1.DOWNLOAD THE DATASET**

The given dataset has been downloaded successfully

**2.LOAD THE DATASET**

**Question-2:**

Load the dataset.

**Solution:**

**2.LOAD THE DATASET**

```
[ ] import numpy as np
```

```
[ ] import pandas as pd
```

```
[ ] df = pd.read_csv("Churn_Modelling.csv")
```

```
[ ] df
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.88	1
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1

```
[ ] df.head()
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.88	1
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
3	4	15701354	Boni	699	France	Female	39	1	0.00	2	0	0	93826.63	0
4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79084.10	0

```
[ ] import numpy as np
```

```
[ ] import pandas as pd
```

```
[ ] import matplotlib.pyplot as plt
```

### Question-3:

Perform Below Visualizations.

#### 3 a) Univariate Analysis

+ Code + Text

Connect Editing

3) Perform Below Visualizations. a) Univariate Analysis

```
[ ] import numpy as np
```

```
[ ] import pandas as pd
```

```
[ ] import matplotlib.pyplot as plt
```

```
[ ] import seaborn as sns
```

```
[ ] data=pd.read_csv("Churn_Modelling.csv")
```

```
data.head()
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.88	1
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1

```
sns.distplot(data['EstimatedSalary'])
```

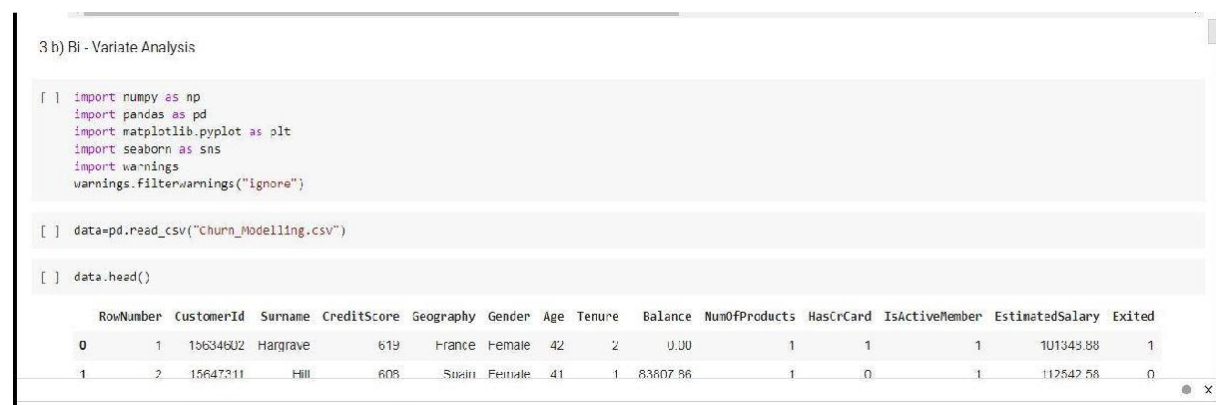
/usr/local/lib/python3.7/dist-packages/seaborn/distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. warnings.warn(msg, FutureWarning)

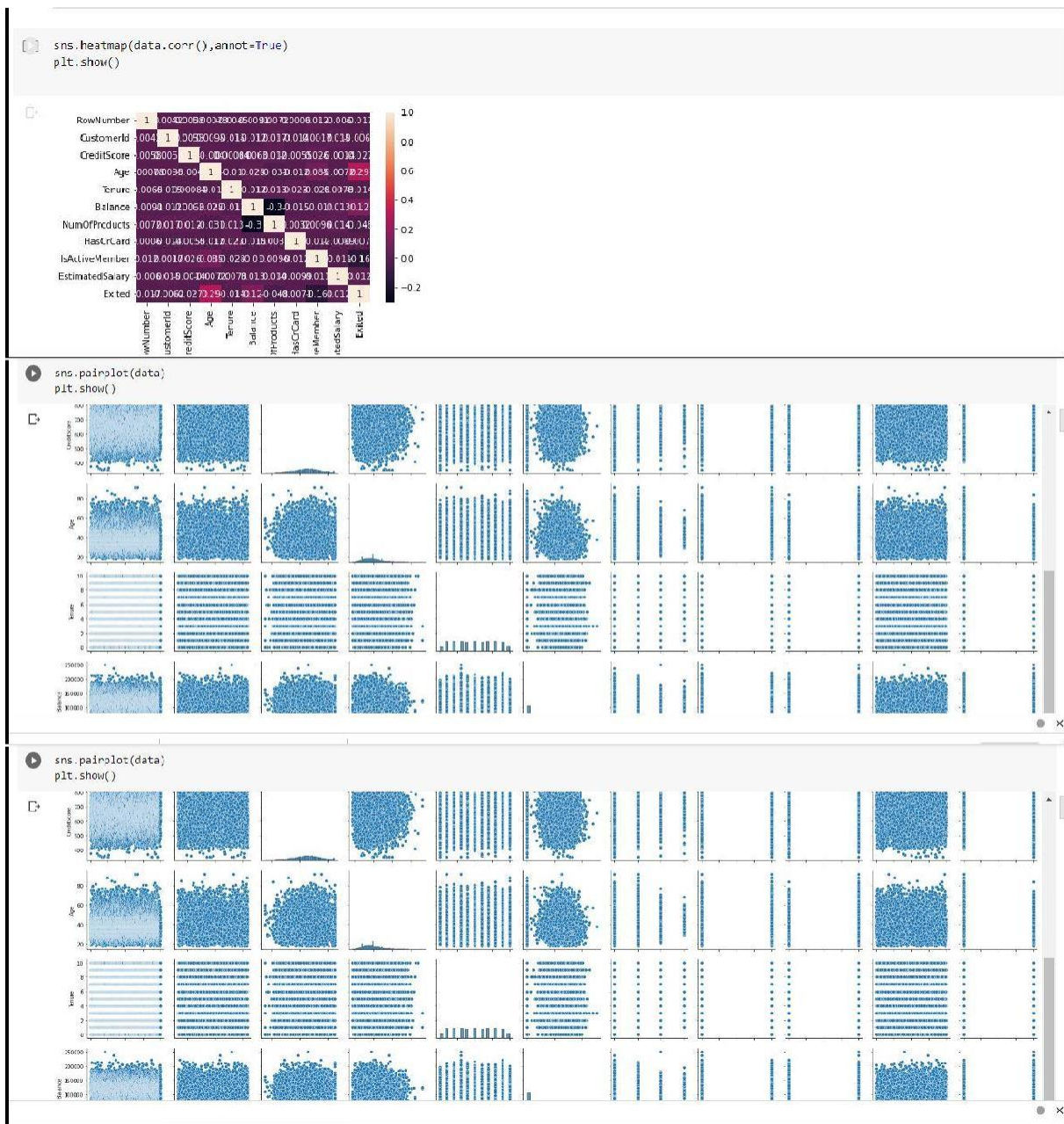
<matplotlib.axes.\_subplots.AxesSubplot at 0x7fed70be7490>





### 3 b) Bi - Variate Analysis





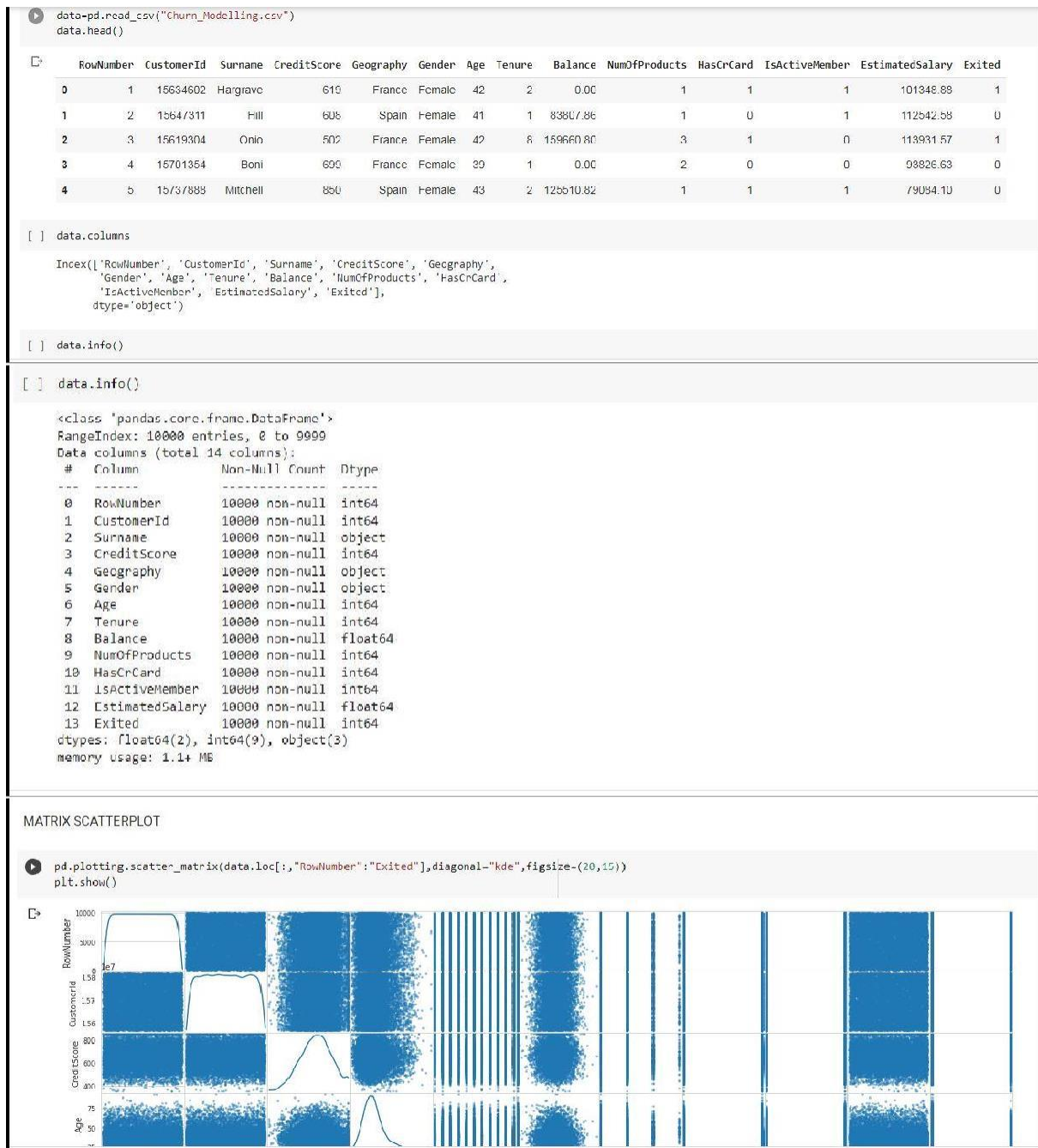
### 3 c) Multi - Variate Analysis

#### 3 C)MULTI-VARIATE ANALYSIS

```
[ ] from pydoc import help
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.preprocessing import scale
from sklearn.decomposition import PCA
from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
from scipy import stats
from IPython.display import display,HTML
%matplotlib inline
np.set_printoptions(suppress=True)
pd.set_option('display.max_rows',20)
import os
print(os.listdir("../NT project/"))
```

```
-----
FileNotFoundError                                Traceback (most recent call last)
<ipython-input-114-7c611291f384> in <module>
```





#### Question-4:

Perform descriptive statistics on the dataset.

##### 4. DESCRIPTIVE STATISTICS

```
[ ] import numpy as np
import pandas as pd
from pandas import Series, DataFrame
import scipy
from scipy import stats
```

```
data=pd.read_csv("Churn_Modelling.csv")
data.head()
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.88	1
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
3	4	15701354	Boni	699	France	Female	39	1	0.00	2	0	0	93826.63	0
4	5	15727688	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79064.10	0

t.

```
data.sum()

RowNumber      50005000
CustomerId      156909405694
Surname      HargraveHillOnioBoniMitchellChuBartlettObinnaH...
CreditScore      6505288
Geography      FranceSpainFranceFranceSpainSpainFranceGermany...
Gender      FemaleFemaleFemaleFemaleFemaleMaleMaleFemaleMa...
Age      389218
Tenure      50128
Balance      764858892.88
NumOfProducts      15302
HasCrCard      7055
IsActiveMember      5151
EstimatedSalary      1000902398.81
Exited      2037
dtype: object
```

```
[ ] data.sum(axis=1)

0      15736618.88
1      15844315.44
2      15803456.37
```

```
[ ] data.median()

RowNumber      5.000500e+03
CustomerId      1.569074e+07
CreditScore      6.520000e+02
Age      3.700000e+01
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
```

```
[ ] data.mean()

RowNumber      5.000500e+03
CustomerId      1.569094e+07
CreditScore      6.505288e+02
Age      3.892180e+01
Tenure      5.012800e+00
```

```
data.max()

RowNumber      10000
CustomerId      15815690
Surname      Zuyeva
CreditScore      850
Geography      Spain
Gender      Male
Age      92
Tenure      10
Balance      250898.09
NumOfProducts      4
HasCrCard      1
IsActiveMember      1
EstimatedSalary      199992.48
Exited      1
dtype: object
```

```
[ ] mpg=data.EstimatedSalary
mpg.idxmax()

6646
```

## LOOKING AT SUMMARY STATISTICS THAT DESCRIBE VARIABLE DISTRIBUTION

```
[ ] data.std()
```

```
RowNumber      2886.895680
CustomerId      71936.186123
CreditScore     96.653299
Age             10.487806
Tenure          2.892174
Balance         62397.405202
NumOfProducts  0.581654
HasCrCard       0.455848
IsActiveMember  0.499797
EstimatedSalary 57510.492818
Exited          0.402769
dtype: float64
```

```
[ ] data.var()
```

```
RowNumber      8.334167e+06
CustomerId      5.174815e+09
CreditScore     9.341860e+03
```

```
[ ] num=data.NumOfProducts
num.value_counts()
```

```
1    5084
2    4590
3     266
4       60
Name: NumOfProducts, dtype: int64
```

```
[ ] data.describe()
```

	RowNumber	CustomerId	CreditScore	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
count	10000.00000	1.000000e+04	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.00000	10000.000000	10000.000000	10000.000000
mean	5000.50000	1.569094e+07	650.528800	38.921800	5.012800	76485.889288	1.530200	0.70550	0.515100	100090.239881	0.203700
std	2886.89568	7.193619e+04	96.653299	10.487806	2.892174	62397.405202	0.581654	0.45584	0.499797	57510.492818	0.402769
min	1.00000	1.556570e+07	350.000000	18.000000	0.000000	0.000000	1.000000	0.00000	0.000000	11.580000	0.000000
25%	2500.75000	1.562853e+07	584.000000	32.000000	3.000000	0.000000	1.000000	0.00000	0.000000	51002.110000	0.000000
50%	5000.50000	1.569094e+07	650.528800	38.921800	5.012800	76485.889288	1.530200	0.70550	0.515100	100090.239881	0.203700
75%	7500.75000	1.574388e+07	724.000000	45.000000	7.000000	119622.519548	2.000000	1.00000	0.615100	130700.781539	0.303700
max	10000.00000	1.596094e+07	850.000000	59.000000	10.000000	125510.82	4.000000	1.00000	1.000000	151600.435894	1.000000

## Question-5:

### Handle the Missing values

#### 5.HANDLE MISSING VALUE

```
[ ] import pandas as pd
```

```
[ ] data=pd.read_csv("Churn_Modelling.csv")
data.head()
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.88	1
1	2	15647311	Hill	608	Spain	Female	41	1	83607.86	1	0	1	112542.58	0
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
3	4	15701354	Boni	699	France	Female	39	1	0.00	2	0	0	93826.63	0
4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79084.10	0

```
[ ] data.shape
```

(10300, 14)

- ▶ `data.isNull()`

[illegible]

10000 rows x 14 columns

```
[ ] data.isnull().sum()
```

RowNumber	0
CustomerId	0
Surname	0
CreditsScore	0
Geography	0
Gender	0
Age	0
Tenure	0
Balance	0
NumOfProducts	0
HasCrCard	0
IsActiveMember	0
EstimatedSalary	0
Exited	0
dtype: int64	

```
[ ] data.isnull().sum().sum()
```

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## FILLING NULL VALUES

```
df = data.fillna(value=0)
df
```

ID	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited	
	0	1	15634602	Hargrave	€19	France	Female	42	2	0.00	1	1	1	101346.88	1
	1	2	15647311	Hill	€08	Spain	Female	41	1	83€07.86	1	0	1	112542.58	0
	2	3	15619304	Onio	€02	France	Female	42	8	159€60.80	3	1	0	113931.57	1
	3	4	15701354	Boni	€99	France	Female	39	1	0.00	2	0	0	93826.63	0
	4	5	15737888	Mitchell	€50	Spain	Female	43	2	125€10.82	1	1	1	79034.10	0
	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
9995	9996	15606229	Objiaiku	771	France	Male	39	5	0.00	2	1	0	96270.64	0	
9996	9997	15669802	Johnstone	€16	France	Male	35	10	57369.61	1	1	1	101690.77	0	
9997	9998	15584532	Liu	709	France	Female	36	7	0.00	1	0	1	42035.58		

10000 rows x 14 columns

```
[1] df.isrull().sum().sum()
```

9

```
[ ] df1=data.fillna(value=5)
df1
```

df1

[illegible]



# FILLING NULL VALUES WITH A PREVIOUS VALUE

```
[ ] df2=data.fillna(method='pad')
df2
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited	
	0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101346.88	1
	1	2	15647311	Hill	603	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
	2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
	3	4	15701354	Bonil	699	France	Female	39	1	0.00	2	0	0	93826.63	0
	4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79034.10	0
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
9995	9996	15606229	Obijaku	771	France	Male	39	5	0.00	2	1	0	96270.64	0	0
9996	9997	15569892	Johnstone	516	France	Male	35	10	57369.61	1	1	1	101696.77	0	0

# FILLING NULL VALUES WITH A PREVIOUS VALUE

```
[ ] df2=data.fillna(method='pad')
df2
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited	
	0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101346.88	1
	1	2	15647311	Hill	603	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
	2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
	3	4	15701354	Boni	699	France	Female	39	1	0.00	2	0	0	93826.63	0
	4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79034.10	0
	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
	9995	9996	15606229	Obijaku	771	France	Male	39	5	0.00	2	1	0	96270.64	0
	9996	9997	15569892	Johnstone	516	France	Male	35	10	57369.61	1	1	1	101696.77	0

```
[ ] df2.isnull().sum().sum()
```

0

```
[ ] #filling NULL values with the next value
df3=data.fillna(method='bfill')
df3
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited	
	0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101346.88	1
	1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
	2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
	3	4	15701354	Boni	599	France	Female	39	1	0.00	2	0	0	93826.63	0
	4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79084.10	0
	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
	9995	9996	15606229	Obijaku	771	France	Male	39	5	0.00	2	1	0	96270.64	0
	9996	9997	15569892	Johnstone	516	France	Male	35	10	57369.61	1	1	1	101690.77	0
	9997	9998	15534532	Liu	709	France	Female	36	7	0.00	1	0	1	42085.58	1

# DROPPING NULL VALUES

```
[ ] df4=data.dropna()
df4
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited	
	0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.88	1
	1	2	15647311	Hill	608	Spain	Female	41	1	83807.36	1	0	1	112542.58	0
	2	3	15619304	Onin	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
	3	4	15701354	Bonil	699	France	Female	39	1	0.00	2	0	0	93826.63	0
	4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79084.10	0
	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
	9995	9996	15606229	Obijaku	771	France	Male	39	5	0.00	2	1	0	96270.64	0
	9996	9997	15569892	Johnstone	516	France	Male	35	10	57369.61	1	1	1	101696.77	0
	9997	9998	15534532	Liu	709	France	Female	36	7	0.00	1	0	1	42085.58	1
	9998	9999	15682355	Sabbatini	772	Germany	Male	42	3	70075.31	2	1	0	92888.52	1
	9999	10000	15628319	Walker	792	France	Female	28	4	130142.79	1	1	0	38190.78	0

<pre>df5=data.dropna(how='any') df5</pre>														
	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.88	1
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
3	4	15701354	Roni	699	France	Female	39	1	0.00	2	0	0	93826.53	0
4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79084.10	0
...														
9995	9996	15606229	Obijaku	771	France	Male	39	5	0.00	2	1	0	96270.54	0
9996	9997	15696892	Johnstone	516	France	Male	35	10	57369.61	1	1	1	101699.77	0
9997	9998	15584532	Liu	709	France	Female	36	7	0.00	1	0	1	42085.58	1
9998	9999	15682355	Sabbatini	772	Germany	Male	42	3	75075.31	2	1	0	92888.52	1
9999	10000	15628319	Walker	782	France	Female	28	4	130142.79	1	1	0	38190.76	0

<pre>replace()</pre>														
	<pre>import numpy as np df6=df.replace(to_replace=np.nan,value=8763) df6</pre>													
	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.83	1
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.53	0
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
3	4	15701354	Roni	699	France	Female	39	1	0.00	2	0	0	93826.53	0
4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79084.10	0
...														
9995	9996	15606229	Obijaku	771	France	Male	39	5	0.00	2	1	0	96270.64	0
9996	9997	15696892	Johnstone	516	France	Male	35	10	57369.61	1	1	1	101699.77	0
9997	9998	15584532	Liu	709	France	Female	36	7	0.00	1	0	1	42085.53	1

<pre>interpolate()</pre>														
	<pre>data['EstimatedSalary']=data['EstimatedSalary'].interpolate(method='linear') data</pre>													
	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.88	1
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
3	4	15701354	Roni	699	France	Female	39	1	0.00	2	0	0	93826.53	0
4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79084.10	0
...														
9995	9996	15606229	Obijaku	771	France	Male	39	5	0.00	2	1	0	96270.54	0
9996	9997	15696892	Johnstone	516	France	Male	35	10	57369.61	1	1	1	101699.77	0
9997	9998	15584532	Liu	709	France	Female	36	7	0.00	1	0	1	42085.50	1
9998	9999	15682355	Sabbatini	772	Germany	Male	42	3	75075.31	2	1	0	92888.52	1
9999	10000	15628319	Walker	782	France	Female	28	4	130142.79	1	1	0	38190.76	0

Question-6:

## Find the outliers and replace the outliers

### 6.FIND THE OUTLIERS AND REPLACE THE OUTLIERS

```
[ ] import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline

[ ] data=pd.read_csv("Churn_Modelling.csv")
data1=data["CreditScore"]
outliers=[]
def detect_outliers(data):
    threshold=3
    mean=np.mean(data)
    std=np.std(data)
    for i in data:
        z_score=(i-mean)/std
        if np.abs(z_score)>threshold:
            outliers.append(z_score)
    return outliers
```

```
[ ] outlier_pt=detect_outliers(data1)
```

+ Code

+ Text

▶ outlier\_pt

```
[ ] outlier_pt=detect_outliers(data1)
```

```
[ ] outlier_pt
```

### INTERQUANTILE RANGE

▶ sorted(data1)

↗

```
351,
358,
359,
363,
365,
367,
373,
376,
376,
382,
383,
386,
395,
399,
401,
404,
405,
521,
521,
521,
521,
521,
521,
521,
521,
521,
521,
521,
521,
...
```

```
[ ] quantile1,quantile3=np.percentile(data1,[25,75])
```

```
[ ] print(quantile1,quantile3)
```

```
584.0 718.0
```

```
[ ] iqr_value=quantile3-quantile1
print(iqr_value)
```

```
134.0
```

```
[ ] lower_bound_val=quantile1-(1.5*iqr_value)
```

```
▶ quantile1,quantile3=np.percentile(data1,[25,75])
```

```
[ ] print(quantile1,quantile3)
```

```
584.0 718.0
```

```
[ ] iqr_value=quantile3-quantile1  
print(iqr_value)
```

```
134.0
```

```
[ ] lower_bound_val=quantile1-(1.5*iqr_value)  
upper_bound_val=quantile3+(1.5*iqr_value)
```

```
[ ] print(lower_bound_val,upper_bound_val)
```

```
383.0 919.0
```

## 7. CHECK FOR CATEGORICAL COLUMNS AND PERFORM ENCODING

### Question-7:

Check for Categorical columns and perform encoding.

## 7. CHECK FOR CATEGORICAL COLUMNS AND PERFORM ENCODING

```
[ ] import pandas as pd  
import numpy as np  
import seaborn as sns  
%matplotlib inline
```

### METHOD I

```
[ ] data=pd.read_csv("Churn_Modelling.csv")  
NEW_DataM1=data  
data1=pd.get_dummies(NEW_DataM1["Gender"])
```

```
[ ] data1.head()
```

	Female	Male
0	1	0
1	1	0

4	1	0
3	1	0
4	1	0

```
NEW_DataM1.drop('Gender',axis='columns')
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	42	2	0.00	1	1	1	101348.88	1
1	2	15647311	Hill	608	Spain	41	1	83807.86	1	0	1	112542.58	0
2	3	15619304	Onio	502	France	42	8	159660.80	3	1	0	113931.57	1
3	4	15701354	Boni	699	France	39	1	0.00	2	0	0	93826.63	0
4	5	15737888	Mitchell	850	Spain	43	2	125510.82	1	1	1	79084.10	0
...	...	...	...	...	...	...	...	...	...	...	...	...	...
9995	9996	15605229	Obijaku	771	France	39	5	0.00	2	1	0	96270.64	0
9996	9997	15569892	Johnstone	516	France	35	10	57369.61	1	1	1	101699.77	0
9997	9998	15584532	Liu	709	France	35	7	0.00	1	0	1	42085.58	1
9998	9999	15682355	Sabbatini	772	Germany	42	3	75075.31	2	1	0	92888.52	1

9999	10000	15628319	Walker	792	France	28	4	130142.79	1	1	0	38190.78	0
------	-------	----------	--------	-----	--------	----	---	-----------	---	---	---	----------	---

10000 rows x 13 columns

+ Code + Text

```
[ ] NEW_DataM1["Male"]=data1["Male"].to_list()
NEW_DataM1["Female"]=data1["Female"].to_list()
```

```
[ ] NEW_DataM1
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.88	1
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
3	4	15701354	Boni	699	France	Female	39	1	0.00	2	0	0	93826.63	0
4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79084.10	0
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
9995	9996	15605229	Obijaku	771	France	Male	39	5	0.00	2	1	0	96270.64	0
9996	9997	15569892	Johnstone	516	France	Male	35	10	57369.61	1	1	1	101699.77	0
9997	9998	15584532	Liu	709	France	Female	35	7	0.00	1	0	1	42085.58	1
9998	9999	15682355	Sabbatini	772	Germany	Male	42	3	75075.31	2	1	0	92888.52	1
9999	10000	15628319	Walker	792	France	Female	28	4	130142.79	1	1	0	38190.78	0

10000 rows x 16 columns

```
[ ] NEW_DataM1.head(2)
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited	Male
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.88	1	0
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	0	0

## METHOD II

```
[ ] from sklearn.preprocessing import LabelEncoder
```

```
[ ] data=pd.read_csv("Churn_Modelling.csv")
le=LabelEncoder()
data1=le.fit_transform(data["Gender"])
```



```

[ ] from sklearn.preprocessing import LabelEncoder

data=pd.read_csv("Churn_Modelling.csv")
le=LabelEncoder()
label=le.fit_transform(data["Gender"])

[ ] le.classes_

array(['Female', 'Male'], dtype=object)

[ ] Data=NEW_DataM1.drop("Gender",axis='columns')
Data

```

	RowNumber	CustomerId	Surname	Creditscore	Geography	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited	Male	Female
0	1	15634602	Hargrave	619	France	42	2	0.00	1	1	1	101348.88	1	0	
1	2	15647311	Hill	608	Spain	41	1	83307.86	1	0	1	112542.58	0	0	
2	3	15619304	Onio	502	France	42	8	159660.80	3	1	0	113931.57	1	0	
3	4	15701354	Boni	699	France	39	1	0.00	2	0	0	93826.63	0	0	
4	5	15737888	Mitchell	850	Spain	43	2	125510.82	1	1	1	79084.10	0	0	
9995	9996	15600229	Obijaku	771	France	35	5	0.00	2	1	0	96270.64	0	1	
9996	9997	15569892	Johnstone	516	France	35	10	57369.61	1	1	1	101699.77	0	1	
9997	9998	15584532	Liu	709	France	36	7	0.00	1	0	1	42085.58	1	0	
9998	9999	15682355	Sabbatini	772	Germany	42	3	75075.31	2	1	0	92888.52	1	1	
9999	10000	15628319	Walker	702	France	28	4	130142.79	1	1	0	38190.78	0	0	

10000 rows x 15 columns

```

[ ] Data["Gender"]=label
Data

```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited	Male	Female
0	1	15634602	Hargrave	619	France	42	2	0.00	1	1	1	101348.88	1	0	
1	2	15647311	Hill	608	Spain	41	1	83307.86	1	0	1	112542.58	0	0	
2	3	15619304	Onio	502	France	42	8	159660.80	3	1	0	113931.57	1	0	
3	4	15701354	Boni	699	France	39	1	0.00	2	0	0	93826.63	0	0	
4	5	15737888	Mitchell	850	Spain	43	2	125510.82	1	1	1	79084.10	0	0	

```

[ ] Data["Gender"]-label
Data

```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited	Male	Female
0	1	15634602	Hargrave	619	France	42	2	0.00	1	1	1	101348.88	1	0	
1	2	15647311	Hill	608	Spain	41	1	83307.86	1	0	1	112542.58	0	0	
2	3	15619304	Onio	502	France	42	8	159660.80	3	1	0	113931.57	1	0	
3	4	15701354	Boni	699	France	39	1	0.00	2	0	0	93826.63	0	0	
4	5	15737888	Mitchell	850	Spain	43	2	125510.82	1	1	1	79084.10	0	0	
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
9995	9996	15600229	Obijaku	771	France	35	5	0.00	2	1	0	96270.64	0	1	
9996	9997	15569892	Johnstone	516	France	35	10	57369.61	1	1	1	101699.77	0	1	
9997	9998	15584532	Liu	709	France	36	7	0.00	1	0	1	42085.58	1	0	
9998	9999	15682355	Sabbatini	772	Germany	42	3	75075.31	2	1	0	92888.52	1	1	
9999	10000	15628319	Walker	702	France	28	4	130142.79	1	1	0	38190.78	0	0	

10000 rows x 16 columns

## Question-8:

Split the data into dependent and independent variables.

## 8.SPLIT THE DATA INTO DEPENDENT AND INDEPENDENT VARIABLES

```
[ ] import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
```

```
[ ] data=pd.read_csv("Churn_Modelling.csv")
```

```
▶ X=data.iloc[:,2:9]
X
```

	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance
0	Hargrave	619	France	Female	42	2	0.00
1	Hill	608	Spain	Female	41	1	83807.86
2	Onio	502	France	Female	42	8	159660.80
3	Boni	699	France	Female	39	1	0.00
4	Mitchell	850	Spain	Female	43	2	125510.82
...	...	...	...	...	...	...	...
9995	Obijiaku	771	France	Male	39	5	0.00
9996	Johnstone	516	France	Male	35	10	57369.61
9997	Liu	709	France	Female	36	7	0.00
9998	Sabbatini	772	Germany	Male	42	3	75075.31
9999	Walker	792	France	Female	28	4	130142.79

10000 rows x 7 columns

```
[ ] Y=data.iloc[:,9]
Y
```

0	1
1	1
2	3
3	2
4	1
...	..
9995	2
9996	1
9997	1
9998	2
9999	1

Question-9:

Scale the independent variables

```
Name: NumOfProducts, Length: 10000, dtype: int64
```

## 9.SCALE THE INDEPENDENT VARIABLES

```
[ ] import numpy as np
import pandas as pd
from pandas import Series,DataFrame
import matplotlib.pyplot as plt
from pylab import rcParams
import seaborn as sb
import scipy
import sklearn
from sklearn import preprocessing
from sklearn.preprocessing import scale
```

```
[ ] %matplotlib inline
rcParams['figure.figsize']=5,4
sb.set_style('whitegrid')
```

Normalizing and transforming features with MinMaxScaler() and fit\_transform()

```
[ ] data=pd.read_csv("Churn_Modelling.csv")
```

Normalizing and transforming features with MinMaxScaler() and fit\_transform()

```
[ ] data=pd.read_csv("Churn_Modelling.csv")
```

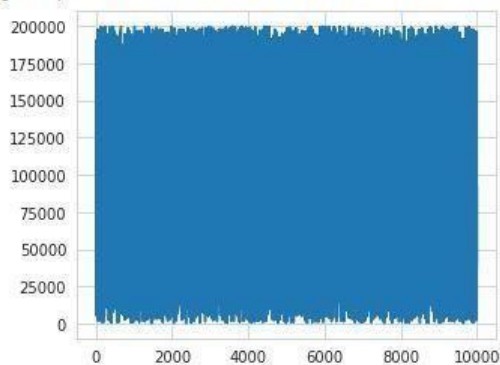
```
[ ] data.head()
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.88	1
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
3	4	15701354	Boni	699	France	Female	39	1	0.00	2	0	0	93826.63	0
4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79084.10	0

```
[ ] tenure=data.EstimatedSalary
plt.plot(tenure)
```

```
[<matplotlib.lines.Line2D at 0x7fed680f7490>]
200000
175000
150000
125000
100000
75000
50000
25000
0
0 2000 4000 6000 8000 10000
plt.plot(tenure)
```

```
[<matplotlib.lines.Line2D at 0x7fed680f7490>]
```



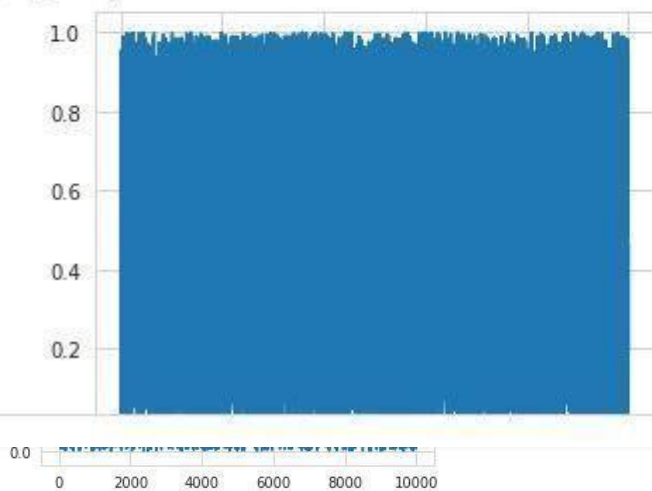
```
[ ] data[['Tenure']].describe()
```

	Tenure
count	10000.000000
mean	5.012800
std	2.892174

min	0.000000
25%	3.000000
50%	5.000000
75%	7.000000
max	10.000000

```
tenure_matrix=tenure.values.reshape(-1,1)
scaled=preprocessing.MinMaxScaler()
scaled_tenure=scaled.fit_transform(tenure_matrix)
plt.plot(scaled_tenure)
```

[<matplotlib.lines.Line2D at 0x7fed680e1750>]

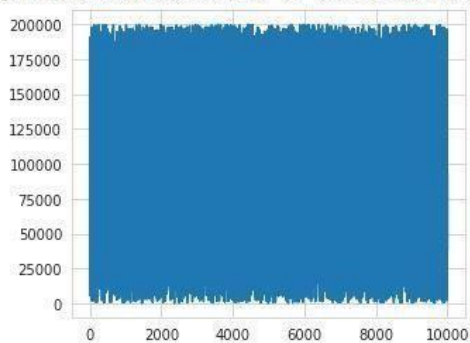


+ Code

+ Text

```
[ ] std_tenure=scale(tenure,axis=0,with_mean=False,with_std=False)
plt.plot(std_tenure)
```

[<matplotlib.lines.Line2D at 0x7fed68046b90>]



+ Code

+ Text

10.SPLIT THE DATA INTO TRAINING AND TESTING

Question-10:



Split the data into training and testing

## 10. SPLIT THE DATA INTO TRAINING AND TESTING

```
[ ] import pandas as pd
data=pd.read_csv("Churn_Modelling.csv")
```

```
data.describe()
```

	RowNumber	CustomerId	CreditScore	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
count	10000.00000	1.000000e+04	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.00000	10000.000000	10000.000000	10000.000000
mean	5000.50000	1.569094e+07	650.528800	38.921800	5.012800	76485.889288	1.530200	0.70550	0.515100	100090.239881	0.203700
std	2886.89568	7.193619e+04	96.653299	10.487806	2.892174	62397.405202	0.581654	0.45684	0.499797	57510.492818	0.402769
min	1.00000	1.556570e+07	350.000000	18.000000	0.000000	0.000000	1.000000	0.00000	0.000000	11.580000	0.000000
25%	2500.75000	1.562853e+07	584.000000	32.000000	3.000000	0.000000	1.000000	0.00000	0.000000	51002.110000	0.000000
50%	5000.50000	1.569074e+07	652.000000	37.000000	5.000000	97198.540000	1.000000	1.00000	1.000000	100193.915000	0.000000
75%	7500.25000	1.575323e+07	718.000000	44.000000	7.000000	127644.240000	2.000000	1.00000	1.000000	149388.247500	0.000000
max	10000.00000	1.581569e+07	850.000000	92.000000	10.000000	250898.090000	4.000000	1.00000	1.000000	199992.480000	1.000000

```
[ ] import numpy as np
```

```
[ ] x=np.array(data["CustomerId"]).reshape(-1,1)
x.shape
```

```
(10000, 1)
```

+ Code

+ Text

```
[ ] y=np.array(data["EstimatedSalary"])
y.shape
```

```
(10000,)
```

```
[ ] print(y)
```

```
[101348.88 112542.58 113931.57 ... 42085.58 92888.52 38190.78]
```

```
[ ] print(type(x))
```

```
<class 'numpy.ndarray'>
```

```
[ ] from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.30)
```

```
[ ] x_train.shape
```

```
(7000, 1)
```

```
[ ] x_test.shape
```

```
(3000, 1)
```

```
[ ] y_train.shape
```

```
(7000,)
```

```
[ ] y.shape
```

```
(10000,)
```

```
[ ] print(y_train.shape)
```

```
(7000,)
```

```
[ ] print(y_test.shape)
```

```
(3000,)
```

