01. Impot libraries

In [63]:
```
import pandas as pd
  import numpy as np
  import matplotlib.pyplot as plt
  import seaborn as sns
```

02. Upload database

In [2]:
```
from google.colab import drive
  drive.mount('/content/drive')
```

```
Mounted at /content/drive
```

Read the Dataset

In [3]:
```
mydata = pd.read_csv('/content/drive/MyDrive/Colab Notebooks/Churn_Modelling.csv'
```

In [8]:
```
mydata.shape
```

Out[8]:
```
(10000,   14)
```

In [6]:
```
mydata.head()
```

Out[6]:

RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
1	15634602	Hargrave	619	France	Female	42	2	0.0	1	2	15647311	Hill	608
2	15647311	Hill	608	Spain	Female	41	1	83807.8	2	3	15619304	Onio	502
3	15619304	Onio	502	France	Female	42	8	159660.8	3	4	15701354	Boni	699
4	15701354	Boni	699	France	Female	39	1	0.0	4	5	15737888	Mitchell	850
5	15737888	Mitchell	850	Spain	Female	43	2	125510.8					

In [11]:
```
mydata.columns
```

Out[11]:
```
Index(['RowNumber', 'CustomerId',
'Surname', 'CreditScore', 'Geography',
'Gender', 'Age', 'Tenure', 'Balance',
'NumOfProducts', 'HasCrCard',
'IsActiveMember', 'EstimatedSalary',
'Exited'],
 dtype='object')
```

In [12]:
```
mydata.tail()
```

Out[12]:

RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Ba	9995	9996	15606229	Obijiaku	771	France	Male	39	5	9996	9997	15569892
9996	15569892	Johnstone	516	France	Male	35	10	573	9997	9998	15584532	Liu	709	France	Female	36	7			

RowNumber CustomerId Surname CreditScore Geography Gender Age Tenure Ba 9998 9999

15682355 Sabbatini 772 Germany Male 42 3 750 **9999** 10000 15628319 Walker 792 France Female 28 4 130

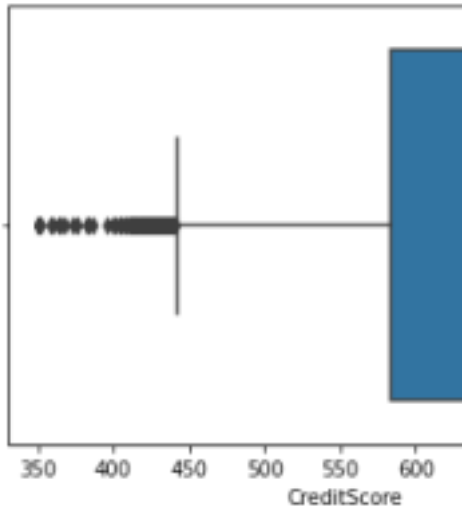
03. Perform Visualizations

In [99]:
```
sns.boxplot(mydata['CreditScore'])
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning:
Pass the following variable as a keyword arg: x. From version 0.12, the only vali
d positional argument will be `data`, and passing other arguments without an expl
icit keyword will result in an error or misinterpretation.
 FutureWarning
```

Out[99]:
<matplotlib.axes._subplots.AxesSu

bplot at 0x7f6c25826090>

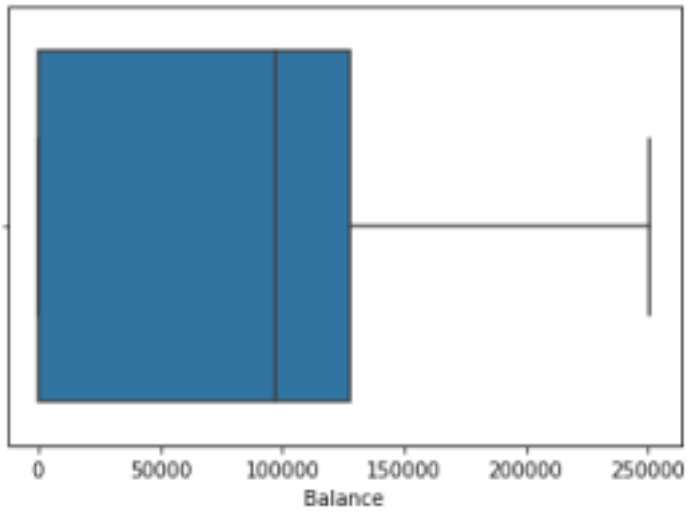


In [102...
```
sns.boxplot(mydata['Balance'])
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning:
Pass the following variable as a keyword arg: x. From version 0.12, the only vali
d positional argument will be `data`, and passing other arguments without an expl
icit keyword will result in an error or misinterpretation.
 FutureWarning
```

Out[102...
<matplotlib.axes._subplots.AxesSu

bplot at 0x7f6c25724310>

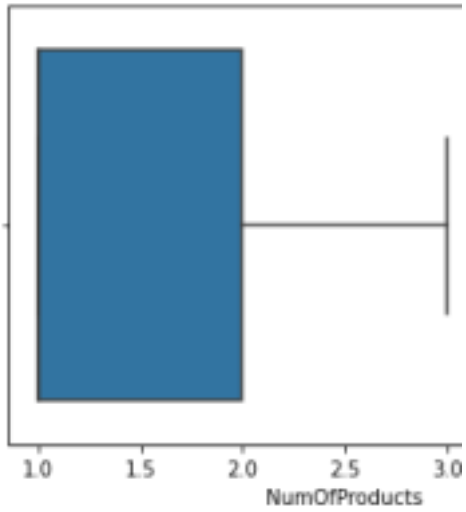


In [103...
```
sns.boxplot(mydata['NumOfProducts'])
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning:
Pass the following variable as a keyword arg: x. From version 0.12, the only vali
d positional argument will be `data`, and passing other arguments without an expl
icit keyword will result in an error or misinterpretation.
 FutureWarning
```

Out[103...
```
<matplotlib.axes._subplots.AxesSu

bplot at 0x7f6c2570ed10>
```

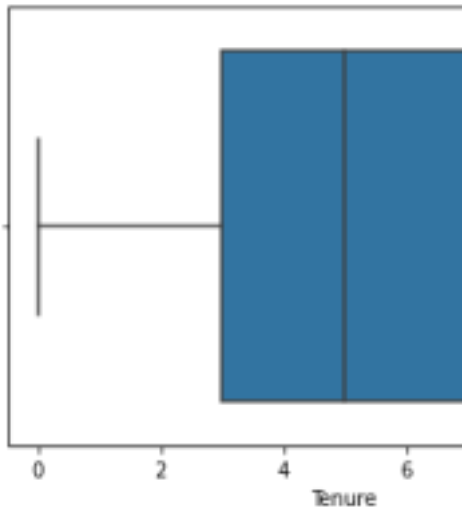


In [104...
```
sns.boxplot(mydata['Tenure'])
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning:
Pass the following variable as a keyword arg: x. From version 0.12, the only vali
d positional argument will be `data`, and passing other arguments without an expl
icit keyword will result in an error or misinterpretation.
 FutureWarning
```

Out[104...
```
<matplotlib.axes._subplots.AxesSu

bplot at 0x7f6c25687550>
```

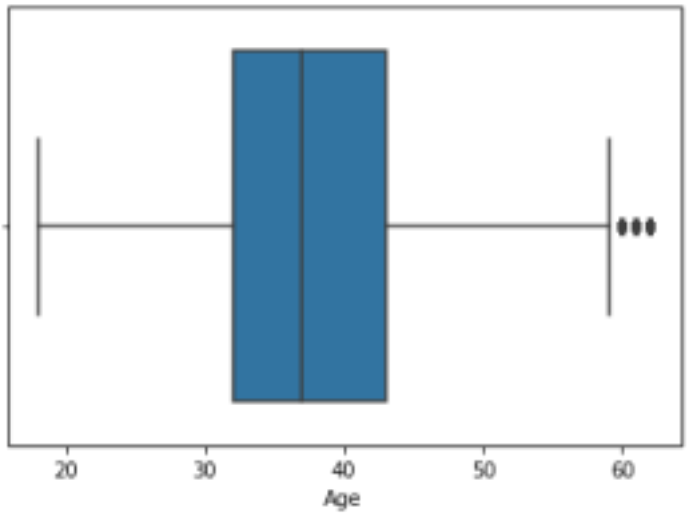


In [105...
```
sns.boxplot(mydata['Age'])
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning:
Pass the following variable as a keyword arg: x. From version 0.12, the only vali
d positional argument will be `data`, and passing other arguments without an expl
icit keyword will result in an error or misinterpretation.
 FutureWarning
```

Out[105...

```
                                bplot at 0x7f6c255fb190>
<matplotlib.axes._subplots.AxesSu
```

In [106...

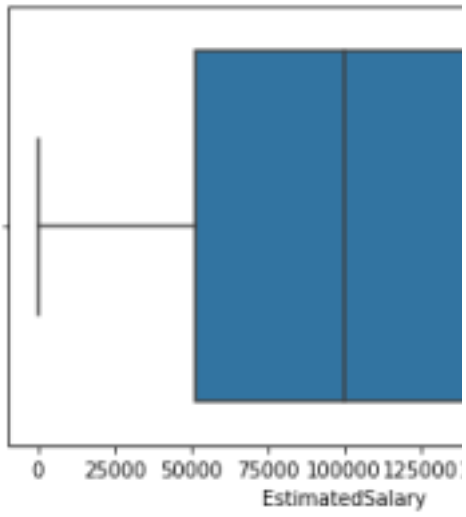
```
sns.boxplot(mydata['EstimatedSalary'])
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning:
Pass the following variable as a keyword arg: x. From version 0.12, the only vali
d positional argument will be `data`, and passing other arguments without an expl
icit keyword will result in an error or misinterpretation.
 FutureWarning
```

Out[106...

```
<matplotlib.axes._subplots.AxesSu

bplot at 0x7f6c255e2450>
```



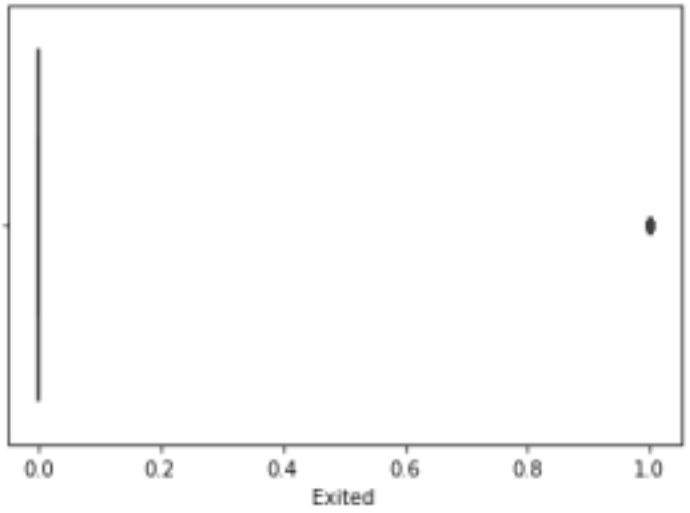
In [107...

```
sns.boxplot(mydata['Exited'])
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning:
Pass the following variable as a keyword arg: x. From version 0.12, the only vali
```

```
d positional argument will be `data`, and passing other arguments without an expl
icit keyword will result in an error or misinterpretation.
 FutureWarning
```

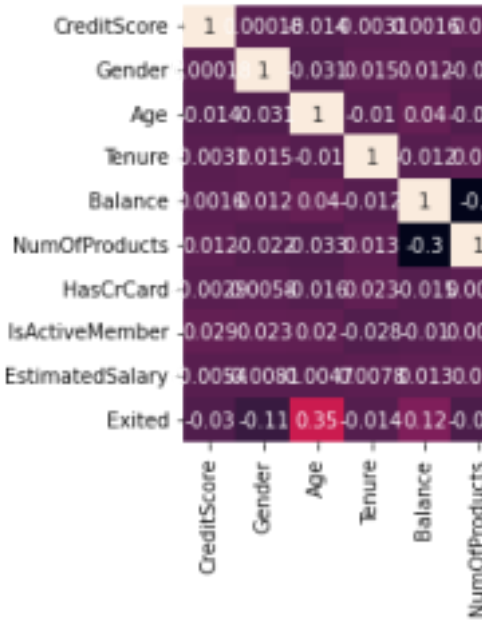
Out[107...
```
                                bplot at 0x7f6c25543890>
<matplotlib.axes._subplots.AxesSu
```

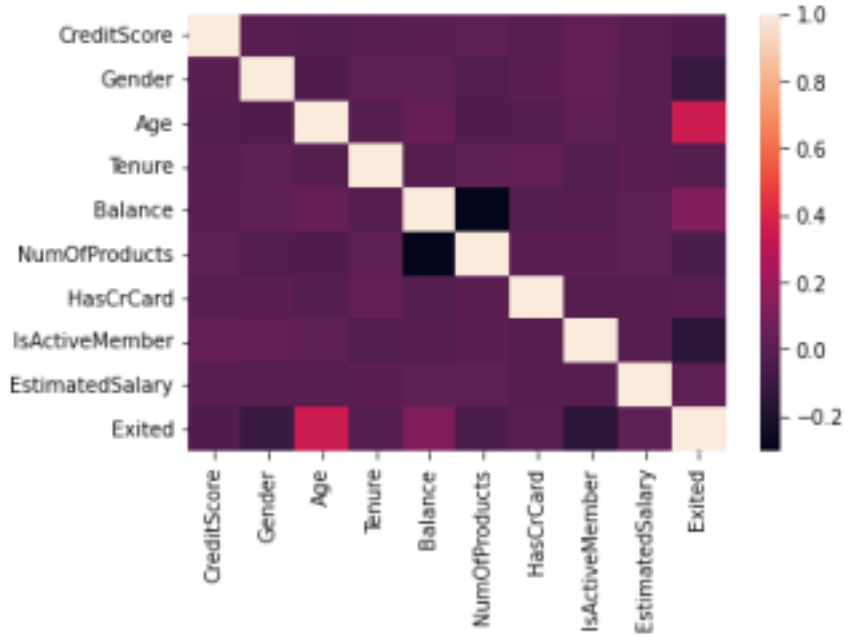
In [108...
```
sns.heatmap(mydata.corr(),annot=True)
```

Out[108...
```
<matplotlib.axes._subplots.AxesSub

plot at 0x7f6c25443f10>
```



```
In [109...  sns.heatmap(mydata.corr(),annot=False)
Out[109...  <matplotlib.axes._subplots.AxesSubplot at 0x7f6c25443750>
```

05. Handling the Missing Values

```
In [16]: mydata.duplicated().sum()
Out[16]: 0
```

```
In [17]: mydata.isna().sum()
Out[17]:
RowNumber 0
CustomerId 0
Surname 0
CreditScore 0
Geography 0
Gender 0 Age 0
Tenure 0
Balance 0
NumOfProducts
0 HasCrCard 0
IsActiveMember
0
EstimatedSalar
```

```
y 0 Exited 0   dtype: int64
```

In [18]:
```
mydata.nunique()
```

Out[18]:
```
                 Tenure 11
                 Balance 6382
                 NumOfProducts 4
RowNumber 10000  HasCrCard 2
CustomerId 10000 IsActiveMember 2
Surname 2932     EstimatedSalary
CreditScore 460  9999
Geography 3
Gender 2 Age 70
```



```
          Exited 2
          dtype: int64
```

In [19]:
```
mydata.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
          dtypes: float64(2), int64(9), object(3)
          memory usage: 1.1+ MB
```

In [25]:
```
mydata.drop(columns=['Gender','HasCrCard','IsActiveMember','Exited']).describe()
```

Out[25]:

	RowNumber	CustomerId	CreditScore	Age	Tenure	Balance	Nu
count	10000.00000	1.000000e+04	10000.000000	10000.000000	10000.000000	10000.000000	
mean	5000.50000	1.569094e+07	650.528800	38.921800	5.012800	76485.889288	
std	2886.89568	7.193619e+04	96.653299	10.487806	2.892174	62397.405202	
min	1.00000	1.556570e+07	350.000000	18.000000	0.000000	0.000000	
25%	2500.75000	1.562853e+07	584.000000	32.000000	3.000000	0.000000	
50%	5000.50000	1.569074e+07	652.000000	37.000000	5.000000	97198.540000	
75%	7500.25000	1.575323e+07	718.000000	44.000000	7.000000	127644.240000	
max	10000.00000	1.581569e+07	850.000000	92.000000	10.000000	250898.090000	

In [26]:
```
qnt=mydata.drop(columns=['Gender','Tenure','HasCrCard','IsActiveMember'])
```

06. Find Outliers

```
In [110]: qnt=mydata.drop(columns=['Gender','Tenure','HasCrCard','IsActiveMember','NumOfPro
          qnt
```

```
Out[110]:
```

	0.0000 51002.1100
0.85	705.0 47.0
	140895.0965 170322.3935

CreditScore Age Balance

EstimatedSalary 0.25 584.0 32.0


```
In [111]: Q1=qnt.iloc[0]
          Q4=qnt.iloc[1]
```

```
In [112]: iqr=Q4-Q1
          iqr
```

```
Out[112]:
```

```
140895.0965
EstimatedSalary
119320.2835 dtype:
float64
```

CreditScore 121.0000
Age 15.0000 Balance

```
In [113]: upper=qnt.iloc[1]+2.5*iqr
          upper
```

```
Out[113]:
```

```
493132.83775
EstimatedSalary
468623.10225 dtype:
float64
```

CreditScore
1007.50000 Age
84.50000 Balance

```
In [114]: lower=qnt.iloc[0]-2.5*iqr
          lower
```

```
Out[114]:
```

```
EstimatedSalary
-247298.59875 dtype:
float64
```

CreditScore
281.50000 Age
-5.50000 Balance
-352237.74125

Replace Outliers

```
In [115]: mydata['CreditScore']= np.where(mydata['CreditScore']>756,650.5288,mydata['Credit
          mydata['Age']= np.where(mydata['Age']>62, 38.9218,mydata['Age'])
```

07. Categorical Columns and Performing Encoding

```
In [132]: mydata['Gender'].replace({'Male': 1, 'Female':0}, inplace=True)
          mydata.head(8)
```

```
Out[132]:
```

CreditScore Gender Age Tenure Balance

NumOfProducts HasCrCard IsActiveMember 0 619.0000

0 42.0 2 0.00 1 1 1 **1** 608.0000 0 41.0 1 83807.86 1 0 1 **2**

502.0000 0 42.0 8 159660.80 3 1 0 **3** 699.0000 0 39.0 1

0.00 2 0 0 **4** 650.5288 0 43.0 2 125510.82 1 1 1 **5** 645.0000

1 44.0 8 113755.78 2 1 0 **6** 650.5288 1 50.0 7 0.00 2 1 1 **7**

376.0000 0 29.0 4 115046.74 4 1 0

Dropping Unwanted Columns

In [144...
```
mydata =mydata.drop(columns=['Age'])
mydata.head()
```

Out[144...

	CreditScore	Gender	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Ex
0	619.0000	0	2	0.00	1	1	1	101348.88	
1	608.0000	0	1	83807.86	1	0	1	112542.58	
2	502.0000	0	8	159660.80	3	1	0	113931.57	
3	699.0000	0	1	0.00	2	0	0	93826.63	
4	650.5288	0	2	125510.82	1	1	1	79084.10	

08. Split Data Into Dependent and Independent Variables

In [78]:
```
a=mydata.iloc[:,:-2]
a.head()
```

Out[78]:

	CreditScore	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember
0	619.0000	0	42.0	2	0.00	1	1	1
1	608.0000	0	41.0	1	83807.86	1	0	1
2	502.0000	0	42.0	8	159660.80	3	1	0
3	699.0000	0	39.0	1	0.00	2	0	0
4	650.5288	0	43.0	2	125510.82	1	1	1

In [83]:
```
b=mydata.iloc[:,-2]
```

In [84]:
```
b.head()
```

Out[84]:
```
0 101348.88
1 112542.58
2 113931.57
3 93826.63
4 79084.10
Name: EstimatedSalary, dtype: float64
```

09. Scale The Independent Variables

In [85]:
```
from sklearn.preprocessing import StandardScaler
```

In [86]:
```
cls=StandardScaler()
a=cls.fit_transform(a)
```

In [130...
```
a
```


Out[130...
```
array([[-0.13284832, -1.09598752,  0.36638802, ..., -0.91158349,
        -1.54776799,  0.97024255],
       [-0.28182929, -1.09598752,  0.48205148, ..., -0.91158349,
         0.64609167,  0.97024255],
       [-1.71746409, -1.09598752,  0.48205148, ...,  2.52705662,
         0.64609167, -1.03067011],
       ...,
```

```
 [ 1.08608688, -1.09598752,
-0.21192932, ..., -0.91158349,
-1.54776799, 0.97024255],
 [ 0.29416906, 0.91241915,
0.48205148, ..., 0.80773656,
0.64609167, -1.03067011],
 [ 0.29416906, -1.09598752,
-1.13723705, ..., -0.91158349,
0.64609167, -1.03067011]])
```

10.Split Data Into Training and Testing

```
In [88]: from sklearn.model_selection import train_test_split

In [92]: a_train,a_test,b_train,b_test= train_test_split(a,b,test_size=0.4,random_state=0)

In [93]: a_train.shape

Out[93]: (6000, 8)

In [94]: a_test.shape
         (4000, 8)
Out[94]:

In [95]: a_train

Out[95]:
array([[-0.67459731, 0.91241915,
0.59771495, ..., -0.91158349,
0.64609167, 0.97024255],
 [ 0.31409458, -1.09598752,
0.25072455, ..., -0.91158349,
0.64609167, 0.97024255],
 [ 0.31409458, 0.91241915,
-0.09626585, ..., 0.80773656,
0.64609167, -1.03067011],
 ...,
 [ 1.47885489, 0.91241915,
-0.32759278, ..., 0.80773656,
0.64609167, -1.03067011],
 [-0.52561634, -1.09598752,
0.01939762, ..., 0.80773656,
0.64609167, 0.97024255],
 [-0.07867343, -1.09598752,
1.17603229, ..., -0.91158349,
0.64609167, -1.03067011]])

In [96]: a_test

Out[96]:
array([[-0.43081026, -1.09598752,
-0.32759278, ..., -0.91158349,
0.64609167, 0.97024255],
 [-1.43304588, -1.09598752,
0.25072455, ..., -0.91158349,
0.64609167, -1.03067011],
 [ 1.04545571, -1.09598752,
0.48205148, ..., -0.91158349,
0.64609167, 0.97024255],
 ...,
 [ 1.58720469, -1.09598752,
-1.59989092, ..., -0.91158349,
-1.54776799, 0.97024255],
 [ 0.84229984, 0.91241915,
0.01939762, ..., 0.80773656,
-1.54776799, -1.03067011],
 [-1.39241471, 0.91241915, -0.90591012, ..., 0.80773656,
-1.54776799, -1.03067011]])

In [131...b_train

Out[131...
7809 198402.37
5279 124550.88
3279 68789.93
8984 42669.37
8466 83343.73
 ...
9225 162961.79
4859 107753.07
3264 181429.87
```

```
9845 148750.16                6000, dtype: float64
2732 118855.26
Name: EstimatedSalary, Length:
```

In [98]:
```
b_test
```

Out[98]: In [ ]:

```
9394 192852.67
898 128702.10
2398 75732.25
5906 89368.59
2343 135662.17
 ...
4758 48545.10
9914 180844.81
7067 94105.00
4578 44653.50
4202 100995.68
Name: EstimatedSalary, Length:
4000, dtype: float64
```