**Assignment -2**

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| --- | --- |
| Assignment Date | 19 September 2022 |
| Student Name | Ajay E |
| Student Roll Number | 210519205003 |
| Maximum Marks | 2 Marks |

**Question 1:**

**Importing Libraries**

import numpy as np

import pandas as pd

import seaborn as sns

import matplotlib.pyplot as plt

**Queastion 2:**

**Loading Data Set**

ds=pd.read\_csv(r'C:\Users\veera\Desktop\Assignment 2/Churn\_Modelling.csv')

ds.shape

(10000, 14)

ds.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 |

**Question 3:**

**Visualizations**

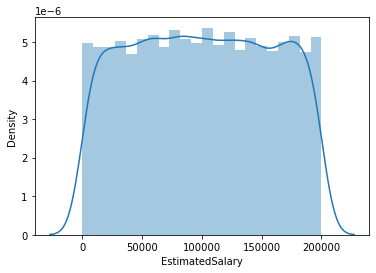
**Univariate**

sns.distplot(ds['EstimatedSalary'],hist=True)

C:\Users\veera\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 flexibility) or `histplot` (an axes-level function for histograms).

  warnings.warn(msg, FutureWarning)

<AxesSubplot:xlabel='EstimatedSalary', ylabel='Density'>

****

## Bivariate

sns.relplot(x='Age',y='Balance',data=ds)

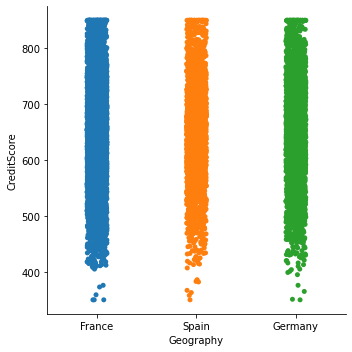
<seaborn.axisgrid.FacetGrid at 0x25cfcd45b20>

## C:\Users\UDHAI\AppData\Local\Microsoft\Windows\INetCache\Content.MSO\BD9F0FAE.tmp

#categorical data

sns.catplot(x='Geography',y='CreditScore',data=ds)

<seaborn.axisgrid.FacetGrid at 0x25cfcbe7490>

****

## Multivariate

sns.relplot(x='Age',y='Balance',hue='Geography',data=ds)

<seaborn.axisgrid.FacetGrid at 0x25cfdd4f310>

## C:\Users\UDHAI\AppData\Local\Microsoft\Windows\INetCache\Content.MSO\3C16B0DA.tmp

#categorical data

sns.boxplot(x='Exited',y='CreditScore',hue='Gender',data=ds)

<AxesSubplot:xlabel='Exited', ylabel='CreditScore'>

## C:\Users\UDHAI\AppData\Local\Microsoft\Windows\INetCache\Content.MSO\18527DF8.tmp

## Question 4:

# Descriptive Statistics

|  | **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 | 1.569074e+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 |

# Handling the missing(null) values

ds.isnull().any()

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

ds.isnull().sum()

RowNumber 0

CustomerId 0

Surname 0

CreditScore 0

Geography 0

Gender 0

Age 0

Tenure 0

Balance 0

NumOfProducts 0

HasCrCard 0

IsActiveMember 0

EstimatedSalary 0

Exited 0

dtype: int64

## Question 5:

# Split the data into dependent and independent variables

x=ds.iloc[:,3:13].values

print(x.shape)

y=ds.iloc[:,13:14].values

print(y.shape)

(10000, 10)

(10000, 1)

**Question 6:**

# Finding and Replacing Outliers

ds.skew()

C:\Users\veera\AppData\Local\Temp\ipykernel\_5256\4252094064.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.

  ds.skew()

RowNumber 0.000000

CustomerId 0.001149

CreditScore -0.071607

Age 1.011320

Tenure 0.010991

Balance -0.141109

NumOfProducts 0.745568

HasCrCard -0.901812

IsActiveMember -0.060437

EstimatedSalary 0.002085

Exited 1.471611

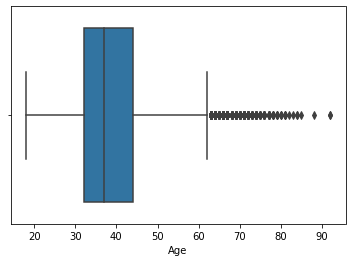
dtype: float64

sns.boxplot(ds["Age"])

C:\Users\veera\anaconda3\lib\site-packages\seaborn\\_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. 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(

<AxesSubplot:xlabel='Age'>

****q0 = ds["Age"].describe()["25%"]

q1 = ds["Age"].describe()["75%"]

iqr=q1-q0

lb = q0 -(1.5\*iqr)

ub = q1 + (1.5\*iqr)

ds[ds["Age"]<lb]

|  | **RowNumber** | **CustomerId** | **Surname** | **CreditScore** | **Geography** | **Gender** | **Age** | **Tenure** | **Balance** | **NumOfProducts** | **HasCrCard** | **IsActiveMember** | **EstimatedSalary** | **Exited** |
| --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- |

ds[ds["Age"]>ub]

|  | **RowNumber** | **CustomerId** | **Surname** | **CreditScore** | **Geography** | **Gender** | **Age** | **Tenure** | **Balance** | **NumOfProducts** | **HasCrCard** | **IsActiveMember** | **EstimatedSalary** | **Exited** |
| --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- |
| **58** | 59 | 15623944 | T'ien | 511 | Spain | Female | 66 | 4 | 0.00 | 1 | 1 | 0 | 1643.11 | 1 |
| **85** | 86 | 15805254 | Ndukaku | 652 | Spain | Female | 75 | 10 | 0.00 | 2 | 1 | 1 | 114675.75 | 0 |
| **104** | 105 | 15804919 | Dunbabin | 670 | Spain | Female | 65 | 1 | 0.00 | 1 | 1 | 1 | 177655.68 | 1 |
| **158** | 159 | 15589975 | Maclean | 646 | France | Female | 73 | 6 | 97259.25 | 1 | 0 | 1 | 104719.66 | 0 |
| **181** | 182 | 15789669 | Hsia | 510 | France | Male | 65 | 2 | 0.00 | 2 | 1 | 1 | 48071.61 | 0 |
| **...** | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| **9753** | 9754 | 15705174 | Chiedozie | 656 | Germany | Male | 68 | 7 | 153545.11 | 1 | 1 | 1 | 186574.68 | 0 |
| **9765** | 9766 | 15777067 | Thomas | 445 | France | Male | 64 | 2 | 136770.67 | 1 | 0 | 1 | 43678.06 | 0 |
| **9832** | 9833 | 15814690 | Chukwujekwu | 595 | Germany | Female | 64 | 2 | 105736.32 | 1 | 1 | 1 | 89935.73 | 1 |
| **9894** | 9895 | 15704795 | Vagin | 521 | France | Female | 77 | 6 | 0.00 | 2 | 1 | 1 | 49054.10 | 0 |
| **9936** | 9937 | 15653037 | Parks | 609 | France | Male | 77 | 1 | 0.00 | 1 | 0 | 1 | 18708.76 | 0 |

359 rows × 14 columns

#Replacing the outlier

outlier\_list = list(ds[ds["Age"] > ub]["Age"])

print(outlier\_list)

[66, 75, 65, 73, 65, 72, 67, 67, 79, 80, 68, 75, 66, 66, 70, 63, 72, 64, 64, 70, 67, 82, 63, 69, 65, 69, 64, 65, 74, 67, 66, 67, 63, 70, 71, 72, 67, 74, 76, 66, 63, 66, 68, 67, 63, 71, 66, 69, 73, 65, 66, 64, 69, 64, 77, 74, 65, 70, 67, 69, 67, 74, 69, 74, 74, 64, 63, 63, 70, 74, 65, 72, 77, 66, 65, 74, 88, 63, 71, 63, 64, 67, 70, 68, 72, 71, 66, 75, 67, 73, 69, 76, 63, 85, 67, 74, 76, 66, 69, 66, 72, 63, 71, 63, 74, 67, 72, 72, 66, 84, 71, 66, 63, 74, 69, 84, 67, 64, 68, 66, 77, 70, 67, 79, 67, 76, 73, 66, 67, 64, 73, 76, 72, 64, 71, 63, 70, 65, 66, 65, 80, 66, 63, 63, 63, 63, 66, 74, 69, 63, 64, 76, 75, 68, 69, 77, 64, 66, 74, 71, 67, 68, 64, 68, 70, 64, 75, 66, 64, 78, 65, 74, 64, 64, 71, 77, 79, 70, 81, 64, 68, 68, 63, 79, 66, 64, 70, 69, 71, 72, 66, 68, 63, 71, 72, 72, 64, 78, 75, 65, 65, 67, 63, 68, 71, 73, 64, 66, 71, 69, 71, 66, 76, 69, 73, 64, 64, 75, 73, 71, 72, 63, 67, 68, 73, 67, 64, 63, 92, 65, 75, 67, 71, 64, 66, 64, 66, 67, 77, 92, 67, 63, 66, 66, 68, 65, 72, 71, 76, 63, 67, 67, 66, 67, 63, 65, 70, 72, 77, 74, 72, 73, 77, 67, 71, 64, 72, 81, 76, 69, 68, 74, 64, 64, 71, 68, 63, 67, 63, 64, 76, 63, 63, 68, 67, 72, 70, 81, 67, 73, 66, 68, 71, 66, 63, 75, 69, 64, 69, 70, 71, 71, 66, 70, 63, 64, 65, 63, 67, 71, 67, 65, 66, 63, 73, 66, 64, 72, 71, 69, 67, 64, 81, 73, 63, 67, 74, 83, 69, 71, 78, 63, 70, 69, 72, 70, 63, 74, 80, 69, 72, 67, 76, 71, 67, 71, 78, 63, 63, 68, 64, 70, 78, 69, 68, 64, 64, 77, 77]

outlier\_dict = {}.fromkeys(outlier\_list,ub)

print(outlier\_dict)

{66: 62.0, 75: 62.0, 65: 62.0, 73: 62.0, 72: 62.0, 67: 62.0, 79: 62.0, 80: 62.0, 68: 62.0, 70: 62.0, 63: 62.0, 64: 62.0, 82: 62.0, 69: 62.0, 74: 62.0, 71: 62.0, 76: 62.0, 77: 62.0, 88: 62.0, 85: 62.0, 84: 62.0, 78: 62.0, 81: 62.0, 92: 62.0, 83: 62.0}

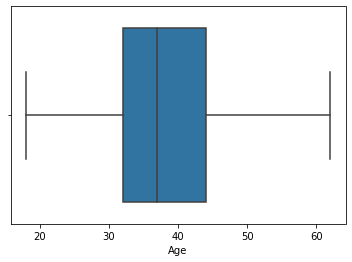
ds["Age"] = ds["Age"].replace(outlier\_dict)

sns.boxplot(ds["Age"])

C:\Users\veera\anaconda3\lib\site-packages\seaborn\\_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. 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(

<AxesSubplot:xlabel='Age'>

****ds[ds["Age"]>ub]

|  | **RowNumber** | **CustomerId** | **Surname** | **CreditScore** | **Geography** | **Gender** | **Age** | **Tenure** | **Balance** | **NumOfProducts** | **HasCrCard** | **IsActiveMember** | **EstimatedSalary** | **Exited** |
| --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- |

**Question 7:**

# Check for Categorical columns and perform encoding

from sklearn.compose import ColumnTransformer

from sklearn.preprocessing import OneHotEncoder

ct=ColumnTransformer([('oh',OneHotEncoder(),[1,2])],remainder='passthrough')

x=ct.fit\_transform(x)

print(x.shape)

(10000, 13)

# saving the data

import joblib

joblib.dump(ct,"churnct.pkl")

['churnct.pkl']

**Question 8:**

# Split the data into training and testing

from sklearn.model\_selection import train\_test\_split

x\_train,x\_test,y\_train,y\_test = train\_test\_split(x,y,test\_size=0.2,random\_state=0)

print(x\_train.shape)

print(x\_test.shape)

(8000, 13)

(2000, 13)

from sklearn.preprocessing import StandardScaler

sc=StandardScaler()

x\_train=sc.fit\_transform(x\_train)

x\_test=sc.transform(x\_test)

joblib.dump(sc,"churnsc.pkl")

['churnsc.pkl']