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| --- | --- |
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* ASSIGNMENT-II

import pandas as pd import seaborn as sns import matplotlib.pyplot as plt import numpy as np sns.set\_style('darkgrid') sns.set(font\_scale=1.3)

df=pd.read\_csv("/content/drive/MyDrive/IBM/Assignment - 2

/Churn\_Modelling.csv") df.head()

RowNumber CustomerId Surname CreditScore Geography Gender Age

\

1. 1 15634602 Hargrave 619 France Female 42
2. 2 15647311 Hill 608 Spain Female 41
3. 3 15619304 Onio 502 France Female 42
4. 4 15701354 Boni 699 France Female 39
5. 5 15737888 Mitchell 850 Spain Female 43

Tenure Balance NumOfProducts HasCrCard IsActiveMember \ 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

df.drop(["RowNumber","CustomerId","Surname"],axis=1,inplace=True) df.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 10000 entries, 0 to 9999

Data columns (total 11 columns):

# Column Non-Null Count Dtype

--- ------ -------------- -----

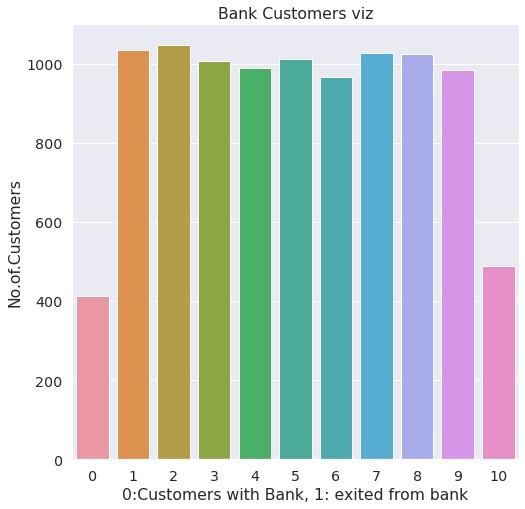
0 CreditScore 10000 non-null int64 1 Geography 10000 non-null object 2 Gender 10000 non-null object 3 Age 10000 non-null int64

4 Tenure 10000 non-null int64 5 Balance 10000 non-null float64 6 NumOfProducts 10000 non-null int64

7 HasCrCard 10000 non-null int64 8 IsActiveMember 10000 non-null int64 9 EstimatedSalary 10000 non-null float64 10 Exited 10000 non-null int64 dtypes: float64(2), int64(7), object(2) memory usage: 859.5+ KB

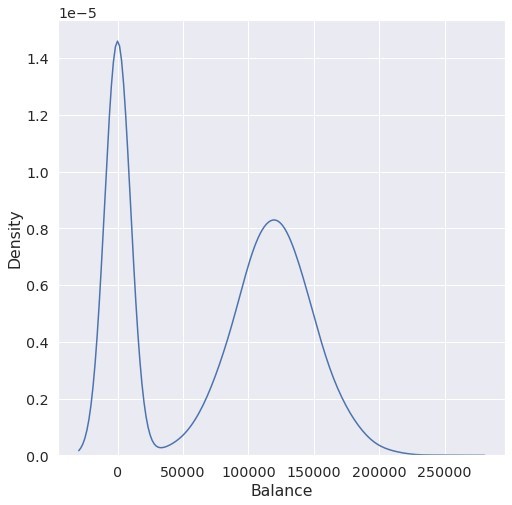
*#Perform Univariate Analysis* plt.figure(figsize=(8,8)) sns.countplot(x='Tenure',data=df)

plt.xlabel('0:Customers with Bank, 1: exited from bank') plt.ylabel('No.of.Customers') plt.title("Bank Customers viz") plt.show()



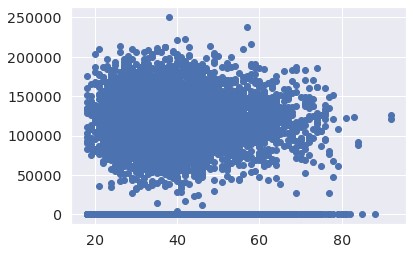
*#Perform Univariate Analysis* plt.figure(figsize=(8,8)) sns.kdeplot(x=df['Balance'])

<matplotlib.axes.\_subplots.AxesSubplot at 0x7fa0c03906d0>



*#Perform Bivariate Analysis* plt.scatter(df.Age,df.Balance)

<matplotlib.collections.PathCollection at 0x7fa0d35a7dd0>



*#Perform Bivariate Analysis* df.corr()

CreditScore Gender Age Tenure

Balance \

CreditScore 1.000000 0.007888 -0.003965 0.000842 0.006268 Gender 0.007888 1.000000 0.022812 0.003739 0.069408

Age -0.003965 0.022812 1.000000 -0.009997 0.028308 Tenure 0.000842 0.003739 -0.009997 1.000000 -0.012254

Balance 0.006268 0.069408 0.028308 -0.012254 1.000000 NumOfProducts 0.012238 0.003972 -0.030680 0.013444 -0.304180

HasCrCard -0.005458 -0.008523 -0.011721 0.022583 -0.014858

IsActiveMember 0.025651 0.006724 0.085472 -0.028362 -0.010084

EstimatedSalary -0.001384 -0.001369 -0.007201 0.007784 0.012797

Exited -0.027094 0.035943 0.285323 -0.014001 0.118533

NumOfProducts HasCrCard IsActiveMember EstimatedSalary \

CreditScore 0.012238 -0.005458 0.025651 -

0.001384

Gender 0.003972 -0.008523 0.006724 -

0.001369

Age -0.030680 -0.011721 0.085472 -

0.007201

Tenure 0.013444 0.022583 -0.028362

0.007784

Balance -0.304180 -0.014858 -0.010084

0.012797

NumOfProducts 1.000000 0.003183 0.009612

0.014204

HasCrCard 0.003183 1.000000 -0.011866 -

0.009933

IsActiveMember 0.009612 -0.011866 1.000000 -

0.011421

EstimatedSalary 0.014204 -0.009933 -0.011421

1.000000

Exited -0.047820 -0.007138 -0.156128

0.012097

Exited

CreditScore -0.027094 Gender 0.035943 Age 0.285323 Tenure -0.014001 Balance 0.118533 NumOfProducts -0.047820

HasCrCard -0.007138

IsActiveMember -0.156128 EstimatedSalary 0.012097 Exited 1.000000

*#Perform Bivariate Analysis* import statsmodels.api as sm

*#define response variable* y = df['CreditScore']

*#define explanatory variable* x = df[['EstimatedSalary']]

*#add constant to predictor variables* x = sm.add\_constant(x)

*#fit linear regression model* model = sm.OLS(y, x).fit()

*#view model summary* print(model.summary()) OLS Regression Results

======================================================================

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Dep. Variable: CreditScore R-squared: 0.000

Model: OLS Adj. R-squared:

-0.000

Method: Least Squares F-statistic:

0.01916

Date: Sat, 24 Sep 2022 Prob (F-statistic): 0.890

Time: 05:06:19 Log-Likelihood:

-59900.

No. Observations: 10000 AIC:

1.198e+05

Df Residuals: 9998 BIC:

1.198e+05

Df Model: 1

Covariance Type: nonrobust

======================================================================

=============

coef std err t P>|t|

[0.025 0.975]

----------------------------------------------------------------------

-------------

const 650.7617 1.940 335.407 0.000

646.958 654.565

EstimatedSalary -2.326e-06 1.68e-05 -0.138 0.890 -3.53e05 3.06e-05

======================================================================

========

Omnibus: 132.939 Durbin-Watson:

2.014

Prob(Omnibus): 0.000 Jarque-Bera (JB):

84.242

Skew: -0.072 Prob(JB):

5.10e-19

Kurtosis: 2.574 Cond. No.

2.32e+05

====================================================================== ========

Notes:

1. Standard Errors assume that the covariance matrix of the errors iscorrectly specified.
2. The condition number is large, 2.32e+05. This might indicate that there are strong multicollinearity or other numerical problems.

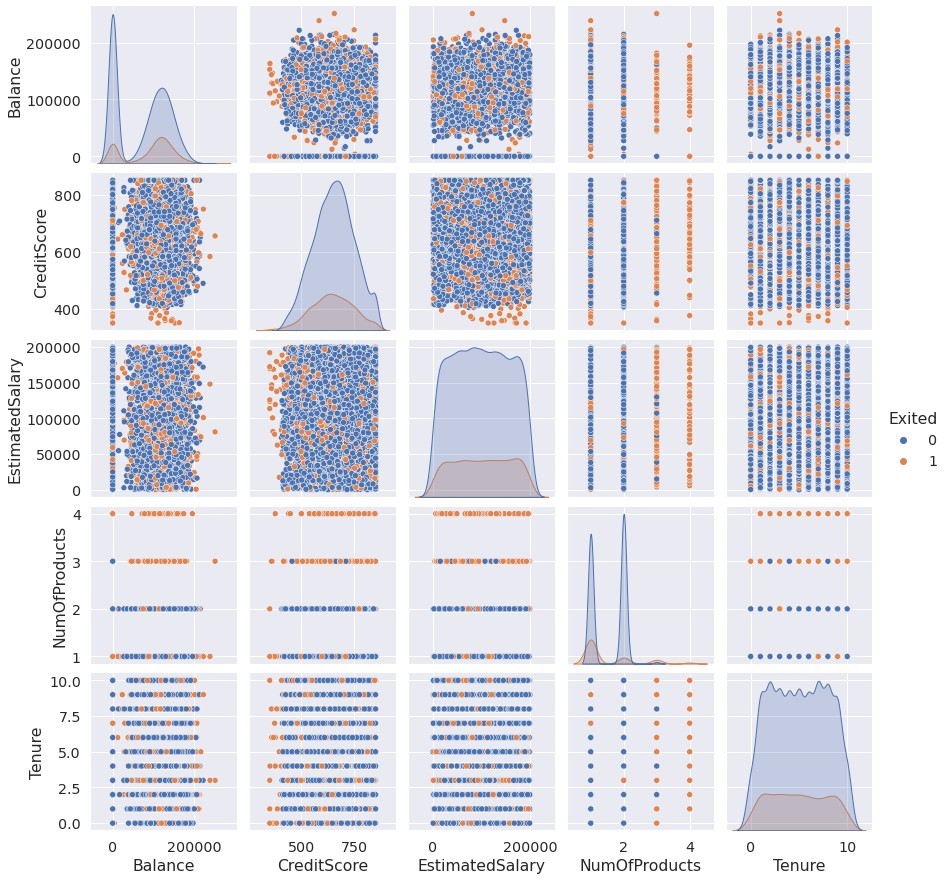
/usr/local/lib/python3.7/dist-packages/statsmodels/tsa/ tsatools.py:142: FutureWarning: In a future version of pandas all arguments of concat except for the argument 'objs' will be keywordonly x = pd.concat(x[::order], 1)

*#Perform Multivariate Analysis* plt.figure(figsize=(4,4))

sns.pairplot(data=df[["Balance","CreditScore","EstimatedSalary","NumOf Products","Tenure","Exited"]],hue="Exited")

<seaborn.axisgrid.PairGrid at 0x7fa0b00a1b10>

<Figure size 288x288 with 0 Axes>



*#Perform Descriptive Statistics* df=pd.DataFrame(df) print(df.sum())

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

*#Perform Descriptive Statistics* print("----Sum Value-----") print(df.sum(1))

print("----------------------------------") print("-----Product Value-----") print(df.prod())

print("----------------------------------")

----Sum Value-----

1. 102015.88
2. 197002.44
3. 274149.37
4. 94567.63
5. 205492.92 ... 9995 97088.64
6. 159633.38
7. 42840.58
8. 168784.83
9. 169159.57

Length: 10000, dtype: float64

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-----Product Value-----

CreditScore 0.0

Age 0.0 Tenure 0.0 Balance 0.0 NumOfProducts 0.0

HasCrCard 0.0

IsActiveMember 0.0 EstimatedSalary inf Exited 0.0 dtype: float64

---------------------------------/usr/local/lib/python3.7/dist-packages/ipykernel\_launcher.py:3:

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.

This is separate from the ipykernel package so we can avoid doing imports until

/usr/local/lib/python3.7/dist-packages/numpy/core/\_methods.py:52: RuntimeWarning: overflow encountered in reduce

return umr\_prod(a, axis, dtype, out, keepdims, initial, where) /usr/local/lib/python3.7/dist-packages/ipykernel\_launcher.py:6:

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.

*#Perform Descriptive Statistics* print("----------Mean Value-----------") print(df.mean())

print("-------------------------------") print("----------Median Value---------") print(df.median())

print("-------------------------------") print("----------Mode Value------------") print(df.mode())

print("-------------------------------")

----------Mean Value-----------

CreditScore 650.528800

Age 38.921800 Tenure 5.012800 Balance 76485.889288 NumOfProducts 1.530200

HasCrCard 0.705500

IsActiveMember 0.515100 EstimatedSalary 100090.239881 Exited 0.203700 dtype: float64

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----------Median Value---------

CreditScore 652.000

Age 37.000 Tenure 5.000 Balance 97198.540 NumOfProducts 1.000

HasCrCard 1.000

IsActiveMember 1.000 EstimatedSalary 100193.915 Exited 0.000 dtype: float64

-------------------------------

----------Mode Value------------

CreditScore Geography Gender Age Tenure Balance

NumOfProducts \

0 850 France Male 37 2 0.0 1

HasCrCard IsActiveMember EstimatedSalary Exited 0 1 1 24924.92 0

-------------------------------

/usr/local/lib/python3.7/dist-packages/ipykernel\_launcher.py:3:

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.

This is separate from the ipykernel package so we can avoid doing imports until

/usr/local/lib/python3.7/dist-packages/ipykernel\_launcher.py:6:

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.

*#Handling with missing Values* df.isnull()*#Checking values are null*

CreditScore Geography Gender Age Tenure Balance

NumOfProducts \

1. False False False False False False

False

1. False False False False False False

False

1. False False False False False False

False

1. False False False False False False

False

1. False False False False False False

False

... ... ... ... ... ... ... ...

1. False False False False False False

False

1. False False False False False False

False

1. False False False False False False

False

1. False False False False False False False
2. False False False False False False False

HasCrCard IsActiveMember EstimatedSalary Exited

1. False False False False
2. False False False False
3. False False False False
4. False False False False
5. False False False False

... ... ... ... ...

1. False False False False
2. False False False False
3. False False False False
4. False False False False
5. False False False False

[10000 rows x 11 columns]

*#Handling with missing Values* df.notnull()*#Checking values are not null*

CreditScore Geography Gender Age Tenure Balance

NumOfProducts \

1. True True True True True True

True

1. True True True True True True

True

1. True True True True True True

True

1. True True True True True True

True

1. True True True True True True

True

... ... ... ... ... ... ... ...

1. True True True True True True

True

1. True True True True True True

True

1. True True True True True True

True

1. True True True True True True

True

1. True True True True True True True

HasCrCard IsActiveMember EstimatedSalary Exited

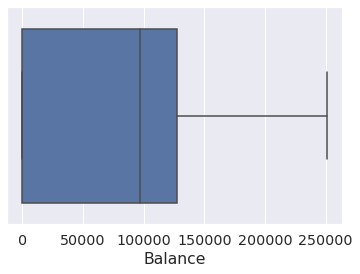
1. True True True True
2. True True True True
3. True True True True
4. True True True True
5. True True True True ... ... ... ... ...
6. True True True True
7. True True True True
8. True True True True
9. True True True True
10. True True True True

[10000 rows x 11 columns]

*#Find outliers & replace the outliers* sns.boxplot(df['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 valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation. FutureWarning

<matplotlib.axes.\_subplots.AxesSubplot at 0x7fa0af6dcf90>



*#Find outliers & replace the outliers* print(np.where(df['Balance']>100000))

(array([ 2, 4, 5, ..., 9987, 9993, 9999]),)

*#Find outliers & replace the outliers* from scipy import stats import numpy as np

z = np.abs(stats.zscore(df["EstimatedSalary"])) print(z)

1. 0.021886
2. 0.216534
3. 0.240687
4. 0.108918
5. 0.365276 ... 9995 0.066419
6. 0.027988
7. 1.008643
8. 0.125231
9. 1.076370

Name: EstimatedSalary, Length: 10000, dtype: float64

*#Check for categorical columns & performs encoding* from sklearn.preprocessing import LabelEncoder df['Gender'].unique()

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

*#Check for categorical columns & performs encoding* df['Gender'].value\_counts()

Male 5457 Female 4543

Name: Gender, dtype: int64

*#Check for categorical columns & performs encoding* encoding=LabelEncoder()

df["Gender"]=encoding.fit\_transform(df.iloc[:,1].values) df

CreditScore Geography Gender Age Tenure Balance

NumOfProducts \

1. 619 France 0 42 2 0.00

1

1. 608 Spain 2 41 1 83807.86

1

1. 502 France 0 42 8 159660.80

3

1. 699 France 0 39 1 0.00

2

1. 850 Spain 2 43 2 125510.82

1

... ... ... ... ... ... ... ...

1. 771 France 0 39 5 0.00

2

1. 516 France 0 35 10 57369.61

1

1. 709 France 0 36 7 0.00

1

1. 772 Germany 1 42 3 75075.31

2

1. 792 France 0 28 4 130142.79 1

HasCrCard IsActiveMember EstimatedSalary Exited

1. 1 1 101348.88 1
2. 0 1 112542.58 0
3. 1 0 113931.57 1
4. 0 0 93826.63 0
5. 1 1 79084.10 0 ... ... ... ... ...
6. 1 0 96270.64 0
7. 1 1 101699.77 0
8. 0 1 42085.58 1
9. 1 0 92888.52 1
10. 1 0 38190.78 0

[10000 rows x 11 columns]

*#Check for categorical columns & performs encoding*

*#Split the data into Dependent & Independent Variables* print("----------Dependent Variables----------")

X=df.iloc[:,1:4] print(X)

print("---------------------------------------") print("---------Independent Variables---------")

Y=df.iloc[:,4] print(Y)

print("---------------------------------------")

----------Dependent Variables-----------

Age Tenure Balance 0 42 2 0.00

1. 41 1 83807.86
2. 42 8 159660.80
3. 39 1 0.00
4. 43 2 125510.82 ... ... ... ... 9995 39 5 0.00
5. 35 10 57369.61
6. 36 7 0.00
7. 42 3 75075.31
8. 28 4 130142.79

[10000 rows x 3 columns]

---------------------------------------

---------Independent Variables---------

1. 1
2. 1
3. 3
4. 2
5. 1 ..
6. 2
7. 1
8. 1
9. 2
10. 1

Name: NumOfProducts, Length: 10000, dtype: int64

---------------------------------------

*#Scale the independent Variables*

from sklearn.preprocessing import StandardScaler object= StandardScaler() *# standardization* scale = object.fit\_transform(df) print(scale)

[[-0.32622142 0.29351742 -1.04175968 ... 0.97024255 0.02188649

1.97716468]

[-0.44003595 0.19816383 -1.38753759 ... 0.97024255 0.21653375

-0.50577476]

[-1.53679418 0.29351742 1.03290776 ... -1.03067011 0.2406869 1.97716468] ...

[ 0.60498839 -0.27860412 0.68712986 ... 0.97024255 -1.00864308

1.97716468]

[ 1.25683526 0.29351742 -0.69598177 ... -1.03067011 -0.12523071

1.97716468]

[ 1.46377078 -1.04143285 -0.35020386 ... -1.03067011 -1.07636976 -0.50577476]]

*#Split the data into training & testing* from sklearn.model\_selection import train\_test\_split

*#Split the data into training & testing*

x\_train, x\_test, y\_train, y\_test = train\_test\_split(x, y, test\_size=4,random\_state=4) x\_train

const EstimatedSalary 2558 1.0 137903.54

7642 1.0 121765.00

8912 1.0 109470.34

3319 1.0 2923.61

6852 1.0 7312.25 ... ... ... 456 1.0 7666.73

6017 1.0 9085.00

709 1.0 147794.63

8366 1.0 102515.42

1146 1.0 54776.64

[9996 rows x 2 columns]

*#Split the data into training & testing* x\_test

const EstimatedSalary 1603 1.0 23305.85

8713 1.0 41248.80

4561 1.0 143317.42

6600 1.0 174123.16

*#Split the data into training & testing* y\_train

2558 727

7642 811

8912 623

3319 430

6852 600 ... 456 733

6017 487

709 686

8366 637

1146 614

Name: CreditScore, Length: 9996, dtype: int64

*#Split the data into training & testing* y\_test

1603 576

8713 786

4561 562

6600 505

Name: CreditScore, dtype: int64