ASSIGNMENT – 2 Python Programming

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Question-1:

1. Importing Required Package

Solution:

```
from google.colab import files
uploaded = files.upload()
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

import warnings
warnings.filterwarnings("ignore")
```

Question-2:

2. Loading the Dataset

Solution:

df=pd.read_csv("/content/Churn_Modelling (1).csv")
df.describe()

Output:

	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

Univariate ananlysis

3. Visualizations

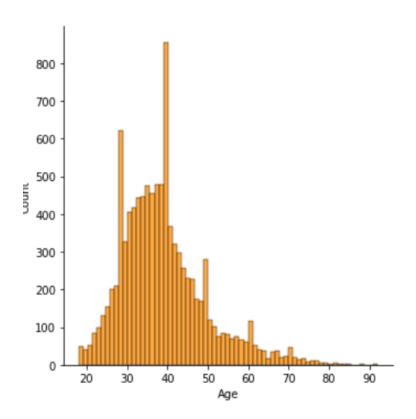
Question-3:

3.1 Univariate Analysis

Solution:

sns.displot(df["Age"], color='darkorange')

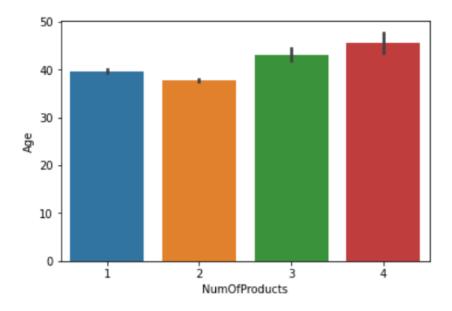
Output:



3.2 Bi-Variate Analysis

Solution:

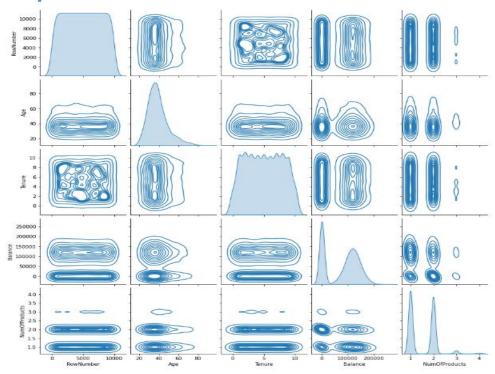
sns.barplot(df["NumOfProducts"],df["Age"])



3.3 Multi - Variate Analysis

Solution:

sns.pairplot(data=df[["RowNumber","Age","Tenure","Balance","NumOfProducts"]],ki
nd="kde")



4. Perform descriptive statistics on the dataset.

Question-4:

Solution:

df.describe()

Output:

	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

5. Handle the Missing values.

Question-5:

Solution:

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

```
D. 0 False
1 False
2 False
3 False
4 False
9995 False
9996 False
9997 False
9998 False
9998 False
9998 False
```

Question-6:

6. Find the outliers and replace the outliers.

Solution:

```
df["Tenure"] = np.where(df["Tenure"] >10, np.median,df["Tenure"])
df["Tenure"]
```

Output:

```
C+ 0 2 1 1 1 2 8 3 3 1 4 4 2 2 9 9995 5 9996 10 9997 7 9998 3 9999 4 Name: Tenure, Length: 1000, dtype: object
```

Question-7:

7. Check for Categorical columns and perform encoding.

Solution:

```
pd.get_dummies(df, columns=["Gender", "Age"], prefix=["Age", "Gender"]
).head()
```

Output:

```
        RowNumber
        CustomerId
        Surname
        CreditScore
        Geography
        Tenure
        Balance
        NumOfProducts
        HasCrCard
        IsActiveMember
        ...
        Gender_78

        0
        1
        15634602
        Hargrave
        619
        France
        2
        0.00
        1
        1
        1
        1
        ...
        0

        1
        2
        15647311
        Hill
        608
        Spain
        1
        83807.86
        1
        0
        1
        ...
        0

        2
        3
        15619304
        Onio
        502
        France
        8
        159660.80
        3
        1
        0
        ...
        0

        3
        4
        15701354
        Boni
        699
        France
        1
        0.00
        2
        0
        0
        ...
        0

        4
        5
        15737888
        Mitchell
        850
        Spain
        2
        125510.82
        1
        1
        1
        1
        ...
        0

        5
        rows × 84 columns
        2
        125510.82
        1
        1
        1
        1
        ...</td
```

```
        HasCrCard
        IsActiveMember
        ...
        Gender_78
        Gender_89
        Gender_81
        Gender_82
        Gender_83
        Gender_84
        Gender_85
        Gender_89
        Gender_92

        1
        1
        ...
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```

Question-8:

- 8. Split the data into dependent and independent variables
- 8.1 Split the data into Independent variables.

Solution:

```
X = df.iloc[:, :-2].values
print(X)
```

Output:

```
☐ [[1 15634602 'Hargrave' ... 1 1 1]
[2 15647311 'Hill' ... 1 0 1]
[3 15619304 'Onio' ... 3 1 0]
...
[9998 15584532 'Liu' ... 1 0 1]
[9999 15682355 'Sabbatini' ... 2 1 0]
[10000 15628319 'Walker' ... 1 1 0]]
```

8.2 Split the data into Dependent variables.

Solution:

```
Y = df.iloc[:, -1].values print(Y)
```

Output:

```
[+ [1 0 1 ... 1 1 0]
```

Question-9:

9. Scale the independent variables

Solution:

```
import pandas as pd
from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler()
df[["RowNumber"]] = scaler.fit_transform(df[["RowNumber"]])
print(df)
```

Question-10:

10. Split the data into training and testing

Solution:

```
from sklearn.model_selection import train_test_split
train_size=0.8
X = df.drop(columns = ['Tenure']).copy()
y = df['Tenure']
X_train, X_rem, y_train, y_rem = train_test_split(X,y, train_size=0.8)
test_size = 0.5
X_valid, X_test, y_valid, y_test = train_test_split(X_rem,y_rem, test_size=0.5)
print(X_train.shape), print(y_train.shape)
print(X_valid.shape), print(y_valid.shape)
print(X_test.shape), print(y_test.shape)
```

```
C* (8000, 13)
(8000,)
(1000, 13)
(1000,)
(1000,)
(1000,)
(None, None)
```