

Data Mining

Project Work

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In [1]: import pandas as pd
 import numpy as np
 import matplotlib.pyplot as plt

#### 1. Read Online Retail Data Set.

[2]:		CLIENTNUM	Attrition_Flag	Customer_Age	Gender	Dependent_count	Education_Le
	0	768805383	Existing Customer	45	М	3	High Sch
	1	818770008	Existing Customer	49	F	5	Gradı
	2	713982108	Existing Customer	51	М	3	Gradı
	3	769911858	Existing Customer	40	F	4	High Scł
	4	709106358	Existing Customer	40	М	3	Uneduca
	•••	•••					
	10122	772366833	Existing Customer	50	М	2	Gradı
	10123	710638233	Attrited Customer	41	М	2	Unkno
•	10124	716506083	Attrited Customer	44	F	1	High Scł
	10125	717406983	Attrited Customer	30	М	2	Gradı
	10126	714337233	Attrited Customer	43	F	2	Gradı
1	0127 rd	ows × 23 colun	nns				
	4						•

### 2.Read First 10 Data.

In [3]: dt.head(5)

Out[3]:		CLIENTNUM	Attrition_Flag	Customer_Age	Gender	Dependent_count	Education_Level
	0	768805383	Existing Customer	45	М	3	High School
	1	818770008	Existing Customer	49	F	5	Graduate
	2	713982108	Existing Customer	51	М	3	Graduate
	3	769911858	Existing Customer	40	F	4	High School
	4	709106358	Existing Customer	40	М	3	Uneducated
	5 rc	ows × 23 colun	nns				



### 3.Read Last 10 Data.

In [4]:	dt.tail(5)								
Out[4]:		CLIENTNUM	Attrition_Flag	Customer_Age	Gender	Dependent_count	Education_Le		
	10122	772366833	Existing Customer	50	М	2	Gradı		
	10123	710638233	Attrited Customer	41	М	2	Unkno		
	10124	716506083	Attrited Customer	44	F	1	High Sch		
	10125	717406983	Attrited Customer	30	М	2	Gradı		
	10126	714337233	Attrited Customer	43	F	2	Gradı		
	5 rows >	< 23 columns							



# 4. Summary of statiscal data

In [5]: dt.describe()

Out[5]:		CLIENTNUM	Customer_Age	Dependent_coun	t Mont	hs_on_book	Total_	Relationship
	count	1.012700e+04	10127.000000	10127.00000	0 1	0127.000000		10127
	mean	7.391776e+08	46.325960	2.34620	3	35.928409		3
	std	3.690378e+07	8.016814	1.29890	8	7.986416		1
	min	7.080821e+08	26.000000	0.00000	0	13.000000		1
	25%	7.130368e+08	41.000000	1.00000	0	31.000000		3
	50%	7.179264e+08	46.000000	2.00000	0	36.000000		4
	75%	7.731435e+08	52.000000	3.00000	0	40.000000		5
	max	8.283431e+08	73.000000	5.00000	0	56.000000		6
	4							•
In [6]:	dt.des	cribe(include=	:'all')					
Out[6]:		CLIENTNUM	Attrition_Flag	Customer_Age	Gender	Dependent_	count	Education_
	count	1.012700e+04	10127	10127.000000	10127	10127.0	00000	
	unique	NaN	2	NaN	2		NaN	
	top	NaN	Existing Customer	NaN	F		NaN	Gra
	freq	NaN	8500	NaN	5358		NaN	
	mean	7.391776e+08	NaN	46.325960	NaN	2.3	46203	
	std	3.690378e+07	NaN	8.016814	NaN	1.2	98908	
	min	7.080821e+08	NaN	26.000000	NaN	0.0	00000	
	25%	7.130368e+08	NaN	41.000000	NaN	1.0	00000	
	50%	7.179264e+08	NaN	46.000000	NaN	2.0	00000	
	75%	7.731435e+08	NaN	52.000000	NaN	3.0	00000	
	max	8.283431e+08	NaN	73.000000	NaN	5.0	00000	
	11 rows	× 23 columns						
	4							•
	<b>.</b>	C						

# 5.Data Types of all the Columns

In [7]: dt.dtypes

```
Out[7]: CLIENTNUM
        int64
        Attrition Flag
        object
        Customer_Age
         int64
        Gender
        object
        Dependent count
         int64
        Education_Level
        object
        Marital Status
        object
        Income_Category
        object
        Card_Category
        object
        Months on book
        int64
        Total_Relationship_Count
        int64
        Months_Inactive_12_mon
        int64
        Contacts_Count_12_mon
        int64
        Credit_Limit
        float64
        Total_Revolving_Bal
        int64
        Avg_Open_To_Buy
        float64
        Total_Amt_Chng_Q4_Q1
        float64
        Total_Trans_Amt
        int64
        Total_Trans_Ct
        int64
        Total_Ct_Chng_Q4_Q1
        float64
        Avg_Utilization_Ratio
        float64
        Naive_Bayes_Classifier_Attrition_Flag_Card_Category_Contacts_Count_12_mon_Dependen
         t_count_Education_Level_Months_Inactive_12_mon_1
                                                             float64
        Naive Bayes Classifier Attrition Flag Card Category Contacts Count 12 mon Dependen
         t_count_Education_Level_Months_Inactive_12_mon_2
                                                             float64
        dtype: object
```

#### 6. Number of Rows

```
In [8]: dt.shape[0]
```

Out[8]: 10127

#### 7. Number of Columns

```
In [9]: dt.shape[1]
Out[9]: 23
```

#### 8.Sum of Any Column

```
In [10]: dt['Dependent_count'].sum()
Out[10]: 23760
```

#### 9. Average Of Any Column

```
In [11]: dt['Dependent_count'].mean()
Out[11]: 2.3462032191172115
```

#### 10. Max in Column

```
In [12]: dt['Dependent_count'].max()
Out[12]: 5
```

#### 11.Min in Columns

```
In [13]: dt['Dependent_count'].min()
Out[13]: 0
```

#### 12. Standard deviation of column

```
In [14]: dt['Dependent_count'].std()
Out[14]: 1.2989083489037916
```

### 13.location of column using iloc

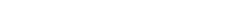
```
In [15]: dt.iloc[3]
```

```
Out[15]: CLIENTNUM
          769911858
         Attrition Flag
          Existing Customer
         Customer_Age
          40
          Gender
         Dependent count
          Education_Level
         High School
         Marital Status
         Unknown
          Income_Category
         Less than $40K
         Card_Category
          Blue
         Months on book
          Total_Relationship_Count
         Months_Inactive_12_mon
         Contacts_Count_12_mon
         Credit_Limit
          3313.0
          Total_Revolving_Bal
          2517
         Avg_Open_To_Buy
          796.0
         Total_Amt_Chng_Q4_Q1
          1.405
         Total_Trans_Amt
         1171
         Total_Trans_Ct
          20
          Total_Ct_Chng_Q4_Q1
          2.333
         Avg_Utilization_Ratio
          0.76
          Naive_Bayes_Classifier_Attrition_Flag_Card_Category_Contacts_Count_12_mon_Dependen
          t_count_Education_Level_Months_Inactive_12_mon_1
                                                                        0.000134
          Naive Bayes Classifier Attrition Flag Card Category Contacts Count 12 mon Dependen
          t_count_Education_Level_Months_Inactive_12_mon_2
                                                                         0.99987
         Name: 3, dtype: object
```

#### **14.copy**

```
In [16]: x=dt.copy()
x
```

Out[16]:		CLIENTNUM	Attrition_Flag	Customer_Age	Gender	Dependent_count	Education_Le
	0	768805383	Existing Customer	45	М	3	High Scł
	1	818770008	Existing Customer	49	F	5	Gradı
	2	713982108	Existing Customer	51	М	3	Gradı
	3	769911858	Existing Customer	40	F	4	High Sch
	4	709106358	Existing Customer	40	М	3	Uneduca
	•••	•••	•••	•••	•••		
	10122	772366833	Existing Customer	50	М	2	Gradı
	10123	710638233	Attrited Customer	41	М	2	Unkno
	10124	716506083	Attrited Customer	44	F	1	High Scł
	10125	717406983	Attrited Customer	30	М	2	Gradı
	10126	714337233	Attrited Customer	43	F	2	Gradı
	10127 rd	ows × 23 colun	nns				



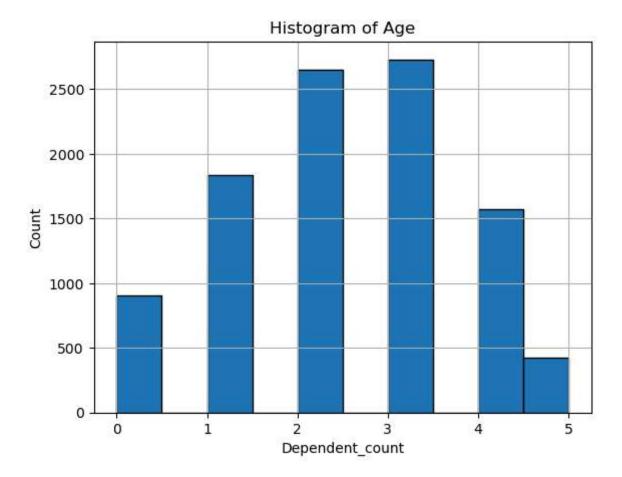
# 15.Uniqu data

```
In [17]: dt['Dependent_count'].nunique()
```

Out[17]: 6

### 16. Histogram of Age

```
In [18]: dt['Dependent_count'].hist(edgecolor = 'black')
    plt.title("Histogram of Age")
    plt.xlabel('Dependent_count')
    plt.ylabel('Count')
    plt.show()
```



### 17.return value having 424 quantity

```
In [19]: q_counts = dt['Dependent_count'].value_counts()
q_counts[q_counts == 424]
```

Out[19]: Dependent\_count 5 424

Name: count, dtype: int64

### 19.drop

```
In [20]: dt.drop("Dependent_count", axis=1, inplace=True)
In [21]: dt
```

CLIENTNUM Attrit					
- CLIENTINOW Attit	tion_Flag	Customer_Age	Gender	Education_Level	Marital_Status
<b>o</b> 768805383	Existing Customer	45	М	High School	Married
<b>1</b> 818770008	Existing Customer	49	F	Graduate	Single
<b>2</b> 713982108	Existing Customer	51	М	Graduate	Married
<b>3</b> 769911858	Existing Customer	40	F	High School	Unknown
<b>4</b> 709106358	Existing Customer	40	М	Uneducated	Married
•••	•••	•••		•••	
<b>10122</b> 772366833	Existing Customer	50	М	Graduate	Single
<b>10123</b> 710638233	Attrited Customer	41	М	Unknown	Divorced
<b>10124</b> 716506083	Attrited Customer	44	F	High School	Married
<b>10125</b> 717406983	Attrited Customer	30	М	Graduate	Unknown
<b>10126</b> 714337233	Attrited Customer	43	F	Graduate	Married
10127 rows × 22 columns					
4					

# 18.Condition

In [22]: dt[dt["Gender"] == 'M']

Out[22]:		CLIENTNUM	Attrition_Flag	Customer_Age	Gender	Education_Level	Marital_Status
	0	768805383	Existing Customer	45	М	High School	Married
	2	713982108	Existing Customer	51	М	Graduate	Married
	4	709106358	Existing Customer	40	М	Uneducated	Married
	5	713061558	Existing Customer	44	М	Graduate	Married
	6	810347208	Existing Customer	51	М	Unknown	Married
	•••	•••	***	•••	•••	•••	•••
	10118	713755458	Attrited Customer	50	М	Unknown	Unknown
	10120	710841183	Existing Customer	54	М	High School	Single
	10122	772366833	Existing Customer	50	М	Graduate	Single
	10123	710638233	Attrited Customer	41	М	Unknown	Divorced
	10125	717406983	Attrited Customer	30	М	Graduate	Unknown
	4769 ro	ws × 22 colum	ns				



# 20.length

```
In [23]: len(dt[dt["Gender"] == 1])
```

Out[23]: 0

# 21.Groupby

```
In [24]: quant = dt.groupby('Gender')['Customer_Age'].sum()
  total = quant[quant > 100]
  total
```

Out[24]: Gender

F 248916 M 220227

Name: Customer\_Age, dtype: int64

#### 22.return Index

#### 25.set Index

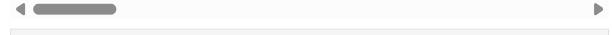
```
In [28]: dt.set_index('CLIENTNUM')
```

> Out[28]: Attrition\_Flag Customer\_Age Gender Education\_Level Marital\_Status Incom **CLIENTNUM** Existing 45 High School Married 768805383 Μ Customer Existing 818770008 49 F Graduate Single Les Customer Existing 713982108 51 Μ Graduate Married Customer Existing 769911858 40 F High School Unknown Les Customer Existing 709106358 40 Μ Uneducated Married Customer Existing 772366833 50 Μ Graduate Single Customer Attrited 710638233 41 Μ Unknown Divorced Customer Attrited F 716506083 44 High School Married Les Customer Attrited 717406983 30 Μ Graduate Unknown Customer Attrited

10127 rows × 21 columns

Customer

714337233



43

F

Graduate

Married

Les