```
In [1]:
```

```
import pandas as pd
```

In [2]:

```
loan=pd.read_excel('loan_sanction.xlsx')
```

In [3]:

loan.head()

Out[3]:

	Unnamed: 0	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncor
0	0	LP001015	Male	Yes	0	Graduate	No	57
1	1	LP001022	Male	Yes	1	Graduate	No	30
2	2	LP001031	Male	Yes	2	Graduate	No	50
3	3	LP001035	Male	Yes	2	Graduate	No	23
4	4	LP001051	Male	No	0	Not Graduate	No	32

In [4]:

loan.tail()

Out[4]:

	Unnamed: 0	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantInc
362	362	LP002971	Male	Yes	3+	Not Graduate	Yes	
363	363	LP002975	Male	Yes	0	Graduate	No	
364	364	LP002980	Male	No	0	Graduate	No	
365	365	LP002986	Male	Yes	0	Graduate	No	
366	366	LP002989	Male	No	0	Graduate	Yes	

In [5]:

loan.columns

Out[5]:

```
In [6]:
```

loan.shape

Out[6]:

(367, 13)

In [7]:

loan=loan.drop(columns='Unnamed: 0') #removing the unnammed column as it was not need loan

Out[7]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Coap
0	LP001015	Male	Yes	0	Graduate	No	5720	
1	LP001022	Male	Yes	1	Graduate	No	3076	
2	LP001031	Male	Yes	2	Graduate	No	5000	
3	LP001035	Male	Yes	2	Graduate	No	2340	
4	LP001051	Male	No	0	Not Graduate	No	3276	
362	LP002971	Male	Yes	3+	Not Graduate	Yes	4009	
363	LP002975	Male	Yes	0	Graduate	No	4158	
364	LP002980	Male	No	0	Graduate	No	3250	
365	LP002986	Male	Yes	0	Graduate	No	5000	
366	LP002989	Male	No	0	Graduate	Yes	9200	

367 rows × 12 columns

In [8]:

```
#checking the duplicates
duplicates = loan[loan.duplicated]
duplicates #seems no duplicate rows
```

Out[8]:

Loan_ID Gender Married Dependents Education Self_Employed ApplicantIncome Coapplica

In [9]:

```
#checking the columns having null values
check_nulls= [col for col in loan.columns if loan[col].isnull().sum()>0]
check_nulls
```

Out[9]:

['Dependents']

In [10]:

loan = loan.dropna() #dropping the rows that have null values

Out[10]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Coap
0	LP001015	Male	Yes	0	Graduate	No	5720	
1	LP001022	Male	Yes	1	Graduate	No	3076	
2	LP001031	Male	Yes	2	Graduate	No	5000	
3	LP001035	Male	Yes	2	Graduate	No	2340	
4	LP001051	Male	No	0	Not Graduate	No	3276	
362	LP002971	Male	Yes	3+	Not Graduate	Yes	4009	
363	LP002975	Male	Yes	0	Graduate	No	4158	
364	LP002980	Male	No	0	Graduate	No	3250	
365	LP002986	Male	Yes	0	Graduate	No	5000	
366	LP002989	Male	No	0	Graduate	Yes	9200	

357 rows × 12 columns

In [11]:

```
#replacing null values with the mode in the gender column
mode_values_gender = loan.mode().iloc[0]
loan = loan.fillna({'Gender' : mode_values_gender['Gender']})
loan
```

Out[11]:

_		Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Coap
-	0	LP001015	Male	Yes	0	Graduate	No	5720	
	1	LP001022	Male	Yes	1	Graduate	No	3076	
	2	LP001031	Male	Yes	2	Graduate	No	5000	
	3	LP001035	Male	Yes	2	Graduate	No	2340	
	4	LP001051	Male	No	0	Not Graduate	No	3276	
	•••								
	362	LP002971	Male	Yes	3+	Not Graduate	Yes	4009	
	363	LP002975	Male	Yes	0	Graduate	No	4158	
	364	LP002980	Male	No	0	Graduate	No	3250	
	365	LP002986	Male	Yes	0	Graduate	No	5000	
	366	LP002989	Male	No	0	Graduate	Yes	9200	

357 rows × 12 columns

In [12]:

```
females = loan[loan['Gender']=='Female']
females
```

Out[12]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Coap
6	LP001055	Female	No	1	Not Graduate	No	2226	
14	LP001096	Female	No	0	Graduate	No	4666	
21	LP001124	Female	No	3+	Not Graduate	No	2083	
23	LP001135	Female	No	0	Not Graduate	No	3765	
30	LP001177	Female	No	0	Not Graduate	No	2478	
332	LP002826	Female	Yes	1	Not Graduate	No	3621	
333	LP002843	Female	Yes	0	Graduate	No	4709	
336	LP002853	Female	No	0	Not Graduate	No	3015	
339	LP002858	Female	No	0	Graduate	No	4333	
360	LP002965	Female	Yes	0	Graduate	No	8550	

68 rows × 12 columns

In [13]:

loan.mode().iloc[0]

Out[13]:

Loan ID	LP001015
Gender	Male
Married	Yes
Dependents	0
Education	Graduate
Self_Employed	No
ApplicantIncome	3500.0
CoapplicantIncome	0.0
LoanAmount	125.0
Loan_Amount_Term	360.0
Credit_History	1.0
Property_Area	Urban
Name: 0, dtype: object	ct

In [14]:

```
#replacing null values with the mode in the gender column
loan = loan.fillna({'Self_Employed' : mode_values_gender['Self_Employed']})
loan
```

Out[14]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Coap
0	LP001015	Male	Yes	0	Graduate	No	5720	
1	LP001022	Male	Yes	1	Graduate	No	3076	
2	LP001031	Male	Yes	2	Graduate	No	5000	
3	LP001035	Male	Yes	2	Graduate	No	2340	
4	LP001051	Male	No	0	Not Graduate	No	3276	
•••								
362	LP002971	Male	Yes	3+	Not Graduate	Yes	4009	
363	LP002975	Male	Yes	0	Graduate	No	4158	
364	LP002980	Male	No	0	Graduate	No	3250	
365	LP002986	Male	Yes	0	Graduate	No	5000	
366	LP002989	Male	No	0	Graduate	Yes	9200	

357 rows × 12 columns

In [15]:

```
#replacing null values with the mode in the gender column
median_loan_amnt = loan['LoanAmount'].median()
median_loan_amnt
loan = loan.fillna({'LoanAmount':median_loan_amnt})
loan
#loan = loan.fillna({''})
```

Out[15]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Coap
0	LP001015	Male	Yes	0	Graduate	No	5720	
1	LP001022	Male	Yes	1	Graduate	No	3076	
2	LP001031	Male	Yes	2	Graduate	No	5000	
3	LP001035	Male	Yes	2	Graduate	No	2340	
4	LP001051	Male	No	0	Not Graduate	No	3276	
•••								
362	LP002971	Male	Yes	3+	Not Graduate	Yes	4009	
363	LP002975	Male	Yes	0	Graduate	No	4158	
364	LP002980	Male	No	0	Graduate	No	3250	
365	LP002986	Male	Yes	0	Graduate	No	5000	
366	LP002989	Male	No	0	Graduate	Yes	9200	

357 rows × 12 columns

In [16]:

```
#top 5 rows
loan.head()
```

Out[16]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Coappli
0	LP001015	Male	Yes	0	Graduate	No	5720	
1	LP001022	Male	Yes	1	Graduate	No	3076	
2	LP001031	Male	Yes	2	Graduate	No	5000	
3	LP001035	Male	Yes	2	Graduate	No	2340	
4	LP001051	Male	No	0	Not Graduate	No	3276	

In [17]:

```
#exporting dataframe
loan.to_excel("new loan min exel.xlsx", index=False)
```

```
In [18]:
```

```
#number of rows and columns
loan.shape
```

Out[18]:

(357, 12)

In [19]:

```
#printi all the columns
loan.columns
```

Out[19]:

In [20]:

```
#descriptive statistics
des_stat = loan.describe()
des_stat
```

Out[20]:

	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History
count	357.000000	357.000000	357.000000	357.000000	357.0
mean	4793.602241	1587.098039	136.501401	343.282913	1.0
std	4936.879103	2357.321156	61.503523	64.015457	0.0
min	0.000000	0.000000	28.000000	6.000000	1.0
25%	2858.000000	0.000000	102.000000	360.000000	1.0
50%	3786.000000	1025.000000	125.000000	360.000000	1.0
75%	5062.000000	2458.000000	158.000000	360.000000	1.0
max	72529.000000	24000.000000	550.000000	480.000000	1.0

In [33]:

```
#Display numeric data type columns in a DF
numeric_cols =[col for col in loan.columns if loan[col].dtype == 'int64']
numeric_cols
```

Out[33]:

```
['ApplicantIncome',
  'CoapplicantIncome',
  'LoanAmount',
  'Loan_Amount_Term',
  'Credit_History']
```

In [26]:

loan.info()

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 357 entries, 0 to 366
Data columns (total 12 columns):
#
     Column
                        Non-Null Count
                                        Dtype
     _____
                        _____
 0
     Loan_ID
                        357 non-null
                                        object
 1
     Gender
                        357 non-null
                                        object
 2
                        357 non-null
     Married
                                        object
 3
     Dependents
                        357 non-null
                                        object
 4
     Education
                        357 non-null
                                        object
 5
     Self Employed
                        357 non-null
                                        object
    ApplicantIncome
 6
                        357 non-null
                                        int64
 7
     CoapplicantIncome 357 non-null
                                        int64
 8
    LoanAmount
                        357 non-null
                                        int64
 9
    Loan_Amount_Term
                        357 non-null
                                        int64
    Credit History
 10
                        357 non-null
                                        int64
    Property Area
                        357 non-null
                                        object
 11
dtypes: int64(5), object(7)
memory usage: 36.3+ KB
```

In [34]:

```
# Select all data types of columns in a DF except object data type.
col_other_than_object = [col for col in loan.columns if loan[col].dtype != 'object'
col_other_than_object
```

Out[34]:

```
['ApplicantIncome',
  'CoapplicantIncome',
  'LoanAmount',
  'Loan_Amount_Term',
  'Credit_History']
```

In [35]:

```
#Extract all the Records where Self_Employed is equal to Yes.
self_employed_yes = loan[loan['Self_Employed']=='Yes']
self_employed_yes
```

)	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome I
ļ	Male	Yes	0	Not Graduate	Yes	2165	3422
)	Male	Yes	0	Graduate	Yes	2267	2792
	Male	No	0	Graduate	Yes	5833	0
)	Male	Yes	3+	Not Graduate	Yes	8000	250
<u>,</u>	Male	Yes	0	Not Graduate	Yes	5293	0
;	Male	Yes	0	Graduate	Yes	7016	292
)	Male	Yes	0	Graduate	Yes	3900	2094
3	Male	No	0	Graduate	Yes	6356	0
;	Male	Yes	0	Graduate	Yes	3188	2286
ŀ	Female	No	0	Graduate	Yes	4742	0
3	Male	Yes	1	Graduate	Yes	3343	1517
	Male	Yes	0	Graduate	Yes	32000	0
ŀ	Male	Yes	2	Graduate	Yes	10890	0
,	Male	No	0	Not Graduate	Yes	8703	0
,	Male	No	0	Not Graduate	Yes	1599	2474
<u> </u>	Male	Yes	2	Graduate	Yes	4246	4246
3	Male	No	0	Graduate	Yes	5833	0
<u> </u>	Male	Yes	0	Graduate	Yes	1900	1442
;	Male	Yes	0	Graduate	Yes	8706	0
?	Male	Yes	3+	Graduate	Yes	5384	0
5	Male	Yes	1	Graduate	Yes	3507	3148
;	Female	No	2	Graduate	Yes	5184	0
3	Male	No	0	Graduate	Yes	2860	2988
3	Male	Yes	2	Graduate	Yes	5000	2166
;	Male	Yes	0	Not Graduate	Yes	3943	0
?	Female	No	0	Graduate	Yes	3333	3916
ŀ	Male	Yes	1	Graduate	Yes	7500	0
,	Male	Yes	1	Not Graduate	Yes	570	2125
	Male	Yes	2	Graduate	Yes	7500	0
3	Male	Yes	3+	Graduate	Yes	6958	1411
,	Male	Yes	1	Graduate	Yes	2360	3355
)	Male	Yes	0	Graduate	Yes	2623	4831

Male No 0 Graduate 3972 4275 Male 0 Graduate 2500 296 Yes Yes Not 4009 1777 Male Yes 3+ Yes Graduate) In Male]: No Graduate Yes 9200 #Extract all the Records where Property area is equal to urban in a DF.

Gender Married Dependents Education Self_Employed ApplicantIncome CoapplicantIncome I

priority_area_urban = loan[loan['Property_Area'] == 'Urban']

Out[36]:

priority_area_urban

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Coap
0	LP001015	Male	Yes	0	Graduate	No	5720	
1	LP001022	Male	Yes	1	Graduate	No	3076	
2	LP001031	Male	Yes	2	Graduate	No	5000	
3	LP001035	Male	Yes	2	Graduate	No	2340	
4	LP001051	Male	No	0	Not Graduate	No	3276	
356	LP002935	Male	Yes	1	Graduate	No	3791	
357	LP002952	Male	No	0	Graduate	No	2500	
360	LP002965	Female	Yes	0	Graduate	No	8550	
362	LP002971	Male	Yes	3+	Not Graduate	Yes	4009	
363	LP002975	Male	Yes	0	Graduate	No	4158	

136 rows × 12 columns

In [40]:

```
#Print number of unique vaues in Gender column
gender_unique_val = loan['Gender'].unique()
print(len(gender_unique_val))
```

2

In []: