```
In [3]: import pandas as pd
import seaborn as sns
import numpy as np
import matplotlib.pyplot as plt
```

```
In [4]: df=pd.read_csv("loan.csv") #read
print(df)
```

	Loan_ID	Gender	Married	Dependents	Educatio	on Self_Employed	
0	LP001002	Male	No	0	Graduat	e No	\
1	LP001003	Male	Yes	1	Graduat	e No	
2	LP001005	Male	Yes	0	Graduat	e Yes	
3	LP001006	Male	Yes	0	Not Graduat	e No	
ļ	LP001008	Male	No	0	Graduat	e No	
					• •	•••	
9	LP002978	Female	No	0	Graduat	e No	
10	LP002979	Male	Yes	3+	Graduat	e No	
11	LP002983	Male	Yes	1	Graduat	e No	
12	LP002984	Male	Yes	2	Graduat	e No	
13	LP002990	Female	No	0	Graduat	ce Yes	
	Applicant	Income	Coappli	cantIncome	LoanAmount	Loan Amount Term	
	• •	5849	• •	0.0	NaN	360.0	\
		4583		1508.0	128.0	360.0	
		3000		0.0	66.0	360.0	
}		2583		2358.0	120.0	360.0	
1		6000		0.0	141.0	360.0	

In [4]: df.describe() #Display descriptive statistics on the dataset

Out[4]:

	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History
count	614.000000	614.000000	592.000000	600.00000	564.000000
mean	5403.459283	1621.245798	146.412162	342.00000	0.842199
std	6109.041673	2926.248369	85.587325	65.12041	0.364878
min	150.000000	0.000000	9.000000	12.00000	0.000000
25%	2877.500000	0.000000	100.000000	360.00000	1.000000
50%	3812.500000	1188.500000	128.000000	360.00000	1.000000
75%	5795.000000	2297.250000	168.000000	360.00000	1.000000
max	81000.000000	41667.000000	700.000000	480.00000	1.000000

In [5]:	df.	head() #	returns	the top	five rows	by defau	Lt		
Out[5]:		Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Coappli
	0	LP001002	Male	No	0	Graduate	No	5849	
	1	LP001003	Male	Yes	1	Graduate	No	4583	
	2	LP001005	Male	Yes	0	Graduate	Yes	3000	
	3	LP001006	Male	Yes	0	Not Graduate	No	2583	
	4	LP001008	Male	No	0	Graduate	No	6000	
	4								+
In [7]:	df[df.isna().any(a	kis=1)]	#Check if	any recor	ds in the dat	a have any mis	sing v
Out[7]:		Loan_I	D Gende	r Married	l Dependents	s Educatio	n Self_Employe	d ApplicantIncom	e Coap
	-	LP00100	2 Mal	e No) () Graduat	e N	o 584	9
	11	LP00102	?7 Mal	e Yes	3	2 Graduat	e Nat	N 250	0
	16	5 LP00103	4 Mal	e No		1 No Graduat		o 359	6
	19	P00104	1 Mal	e Yes	;) Graduat	e Naf	N 260	0
	23	B LP00105	i0 Nal	N Yes	; :	2 No Graduat		0 336	5
	592	2 LP00293	3 Nal	N No	3-	+ Graduat	e Ye	s 935	7
	597	7 LP00294	3 Mal	e No	o Nal	N Graduat	e N	o 298	7
	600	L P00294	9 Femal	e No) 3·			N 41	6
	601	I LP00295	0 Mal	e Yes	S (O Graduat		N 289	4
	605	5 LP00296	60 Mal	e Yes	; (O Graduat	N	o 240	0
	134 rows × 13 columns								
	4								>
In [8]:	df.	dropna(i	nplace=1	True)					
In [9]:	df[df.isna().any(ax	kis=1)]					
Out[9]:	L	.oan_ID G	ender M	arried De	pendents E	ducation S	elf_Employed A	pplicantIncome C	oapplica
	4)

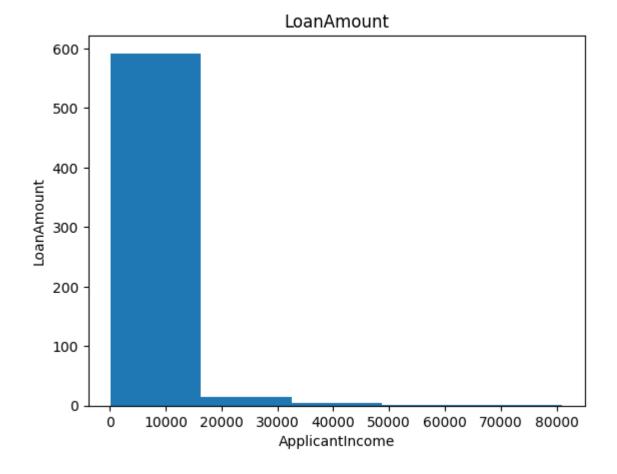
In [10]: pip install matplotlib

Collecting matplotlibNote: you may need to restart the kernel to use updated packages.

```
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                                      460.8/470.9 kB 5.8 MB/s eta 0:00:
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Collecting cycler>=0.10 (from matplotlib)
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Requirement already satisfied: fonttools>=4.22.0 in c:\user\user\appdata\loc
al\programs\python\python311\lib\site-packages (from matplotlib) (4.39.4)
Collecting kiwisolver>=1.0.1 (from matplotlib)
  Downloading kiwisolver-1.4.4-cp311-cp311-win amd64.whl (55 kB)
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grams\python\python311\lib\site-packages (from matplotlib) (1.24.3)
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\programs\python\python311\lib\site-packages (from matplotlib) (23.0)
Requirement already satisfied: pillow>=6.2.0 in c:\user\appdata\local\p
rograms\python\python311\lib\site-packages (from matplotlib) (9.5.0)
Collecting pyparsing>=2.3.1 (from matplotlib)
  Downloading pyparsing-3.0.9-py3-none-any.whl (98 kB)
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          ------ 98.3/98.3 kB 1.4 MB/s eta 0:00:
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Requirement already satisfied: python-dateutil>=2.7 in c:\users\user\appdata
\local\programs\python\python311\lib\site-packages (from matplotlib) (2.8.2)
Requirement already satisfied: six>=1.5 in c:\user\user\appdata\local\progra
ms\python\python311\lib\site-packages (from python-dateutil>=2.7->matplotlib)
(1.16.0)
Installing collected packages: pyparsing, kiwisolver, cycler, contourpy, matp
Successfully installed contourpy-1.1.0 cycler-0.11.0 kiwisolver-1.4.4 matplot
lib-3.7.1 pyparsing-3.0.9
```

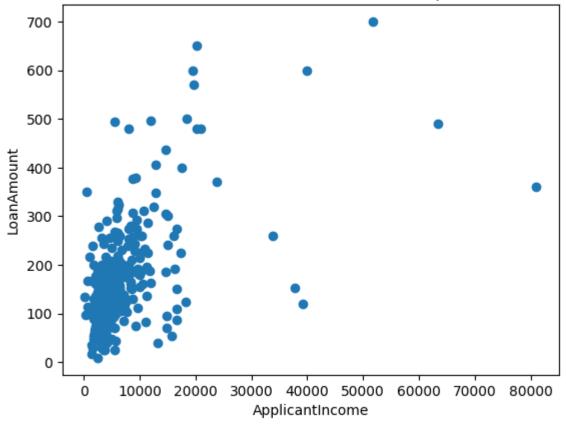
```
In [13]: import matplotlib.pyplot as plt #Build a graph visualizing the distribution of
import pandas as pd
    df=pd.read_csv("loan.csv")
    plt.hist(df['ApplicantIncome'], bins=5)
    plt.xlabel('ApplicantIncome')
    plt.ylabel('LoanAmount')
    plt.title('LoanAmount')
    plt.show()
```



```
In [15]: x=df["ApplicantIncome"].to_numpy()
    y=df["LoanAmount"].to_numpy()
    df1=plt.scatter(x,y)
    plt.title("relation visualization with scatter plot")
    plt.xlabel("ApplicantIncome")
    plt.ylabel("LoanAmount")
```

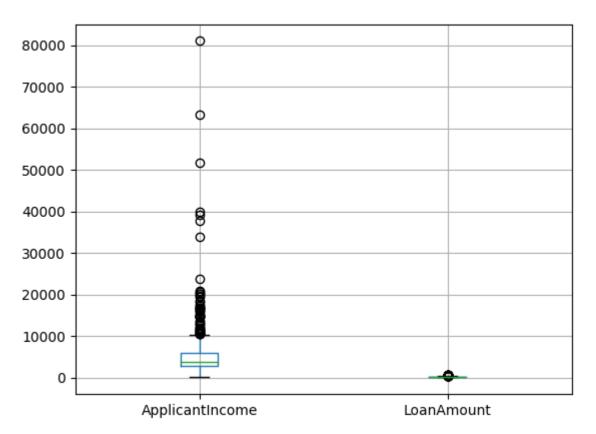
Out[15]: Text(0, 0.5, 'LoanAmount')



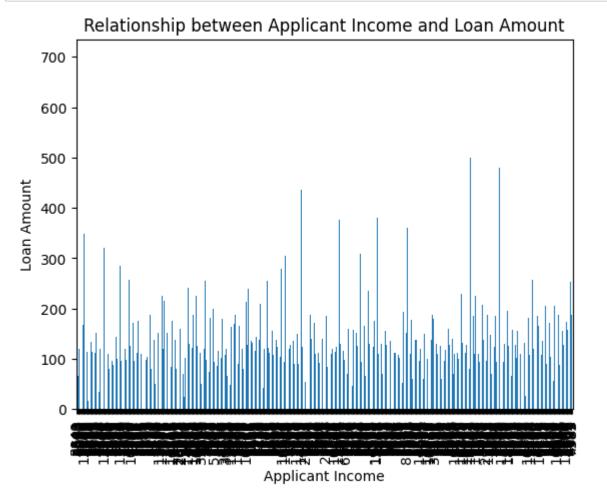


In [19]: df.boxplot(column=['ApplicantIncome', 'LoanAmount']) #Build a graph visualizing

Out[19]: <Axes: >



```
In [6]: df.plot(kind='bar', x='ApplicantIncome', y='LoanAmount', legend=False) #withou
plt.xlabel('Applicant Income')
plt.ylabel('Loan Amount')
plt.title('Relationship between Applicant Income and Loan Amount')
plt.show()
```





In [16]: pip install seaborn

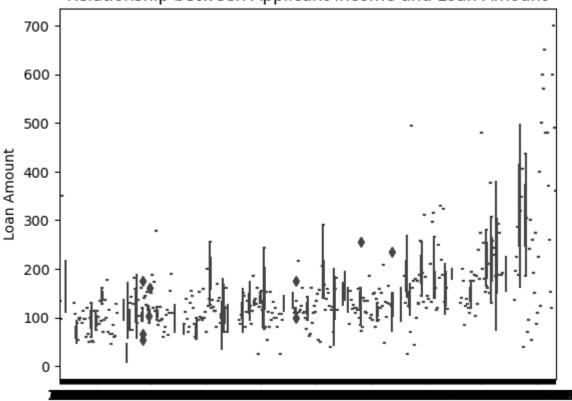
Collecting seabornNote: you may need to restart the kernel to use updated packages.

Downloading seaborn-0.12.2-py3-none-any.whl (293 kB) 0.0/293.3 kB ? eta -:--:--204.8/293.3 kB 6.3 MB/s eta 0:00: 01 286.7/293.3 kB 3.5 MB/s eta 0:00: 01 ----- 293.3/293.3 kB 2.0 MB/s eta 0:00: 00 Requirement already satisfied: numpy!=1.24.0,>=1.17 in c:\users\user\appdata \local\programs\python\python311\lib\site-packages (from seaborn) (1.24.3) Requirement already satisfied: pandas>=0.25 in c:\users\user\appdata\local\pr ograms\python\python311\lib\site-packages (from seaborn) (2.0.1) Requirement already satisfied: matplotlib!=3.6.1,>=3.1 in c:\users\user\appda ta\local\programs\python\python311\lib\site-packages (from seaborn) (3.7.1) Requirement already satisfied: contourpy>=1.0.1 in c:\users\user\appdata\loca l\programs\python\python311\lib\site-packages (from matplotlib!=3.6.1,>=3.1-> seaborn) (1.1.0) Requirement already satisfied: cycler>=0.10 in c:\users\user\appdata\local\pr ograms\python\python311\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seab orn) (0.11.0) Requirement already satisfied: fonttools>=4.22.0 in c:\user\user\appdata\loc al\programs\python\python311\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (4.39.4) Requirement already satisfied: kiwisolver>=1.0.1 in c:\users\user\appdata\loc al\programs\python\python311\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (1.4.4) Requirement already satisfied: packaging>=20.0 in c:\user\user\appdata\local \programs\python\python311\lib\site-packages (from matplotlib!=3.6.1,>=3.1->s eaborn) (23.0) Requirement already satisfied: pillow>=6.2.0 in c:\user\appdata\local\p rograms\python\python311\lib\site-packages (from matplotlib!=3.6.1,>=3.1->sea born) (9.5.0) Requirement already satisfied: pyparsing>=2.3.1 in c:\users\user\appdata\loca l\programs\python\python311\lib\site-packages (from matplotlib!=3.6.1,>=3.1-> seaborn) (3.0.9) Requirement already satisfied: python-dateutil>=2.7 in c:\users\user\appdata \local\programs\python\python311\lib\site-packages (from matplotlib!=3.6.1,>= 3.1->seaborn) (2.8.2) Requirement already satisfied: pytz>=2020.1 in c:\users\user\appdata\local\pr ograms\python\python311\lib\site-packages (from pandas>=0.25->seaborn) (2023. 3) Requirement already satisfied: tzdata>=2022.1 in c:\users\user\appdata\local \programs\python\python311\lib\site-packages (from pandas>=0.25->seaborn) (20 23.3) Requirement already satisfied: six>=1.5 in c:\user\user\appdata\local\progra ms\python\python311\lib\site-packages (from python-dateutil>=2.7->matplotlib! =3.6.1,>=3.1->seaborn) (1.16.0) Installing collected packages: seaborn Successfully installed seaborn-0.12.2

```
In [17]: import seaborn as sns
```

```
In [18]: sns.boxplot(x=df['ApplicantIncome'], y=df['LoanAmount']) #using seaborn
    plt.xlabel('Applicant Income')
    plt.ylabel('Loan Amount')
    plt.title('Relationship between Applicant Income and Loan Amount')
    plt.show()
```





Applicant Income

```
In [19]: unique_values = df['Property_Area'].unique() #Display unique values of a categorint(unique_values)
    ['Urban' 'Rural' 'Semiurban']
```

In [20]: unique_values = df['Education'].unique() #Display unique values of a categoric
print(unique_values)

['Graduate' 'Not Graduate']

```
In [24]: pip install scipy
         Collecting scipy
           Downloading scipy-1.10.1-cp311-cp311-win amd64.whl (42.2 MB)
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         0:21
In [22]: import scipy.stats as stats
In [24]: # Build a contingency table of two potentially related categorical variables.
         contingency_table = pd.crosstab(df['Property_Area'], df['Self_Employed'])
         print(contingency table)
         # Perform the chi-square test of independence
         chi2, p_value, _, _ = stats.chi2_contingency(contingency_table)
         print('Chi-square statistic:', chi2)
         print('p-value:', p_value)
         Self Employed
                         No Yes
         Property_Area
         Rural
                        143
                              26
         Semiurban
                        191
                              32
         Urban
                        166
                              24
         Chi-square statistic: 0.5803134707599138
         p-value: 0.7481462973944264
```

```
In [25]:
         # Retrieve one or more subset of rows based on two or more criteria and present
         subset1 = df[(df['Gender'] == 'Male') & (df['Married'] == 'Yes')]
         subset2 = df[(df['Loan Status'] == 'N') & (df['Property Area'] == 'Rural')]
         # Present descriptive statistics on the subset(s)
         subset1 stats = subset1.describe()
         subset2 stats = subset2.describe()
         print("Subset 1 descriptive statistics:")
         print(subset1_stats)
         print("\nSubset 2 descriptive statistics:")
         print(subset2_stats)
         Subset 1 descriptive statistics:
                 ApplicantIncome CoapplicantIncome
                                                                   Loan Amount Term
                                                      LoanAmount
         count
                      357.000000
                                          357.000000
                                                      343.000000
                                                                         348.000000
                     5529.540616
                                         1828.330308
                                                      154.011662
                                                                         335.931034
         mean
         std
                     6743.209021
                                         2096.367198
                                                       83.025254
                                                                          67.342095
         min
                      150.000000
                                            0.000000
                                                       17.000000
                                                                          12.000000
         25%
                     2882.000000
                                            0.000000
                                                      108.000000
                                                                         360.000000
         50%
                     3875.000000
                                         1619.000000
                                                      132.000000
                                                                         360.000000
         75%
                     5829.000000
                                         2500.000000
                                                      180.000000
                                                                         360.000000
                    81000.000000
                                        20000.000000
                                                      600.000000
                                                                         480.000000
         max
                 Credit History
                     326.000000
         count
                       0.846626
         mean
         std
                       0.360902
         min
                       0.000000
         25%
                       1.000000
         50%
                       1.000000
         75%
                       1.000000
         max
                       1.000000
         Subset 2 descriptive statistics:
                 ApplicantIncome CoapplicantIncome
                                                      LoanAmount
                                                                   Loan Amount Term
                                                                          67.000000
         count
                       69.000000
                                           69.000000
                                                       66.000000
                                                                                      \
         mean
                     6497.521739
                                         1436.507246
                                                      158.742424
                                                                         345.134328
         std
                     9949.733021
                                         1618.225813
                                                       96.412305
                                                                          56.751691
                                            0.000000
                                                       46.000000
                                                                          84,000000
         min
                      150.000000
         25%
                     3365.000000
                                            0.000000
                                                      110.500000
                                                                         360.000000
         50%
                     4283.000000
                                         1287.000000
                                                      133.000000
                                                                         360.000000
         75%
                     6216.000000
                                         2200.000000
                                                      172.750000
                                                                         360.000000
         max
                    81000.000000
                                         5302.000000
                                                      570.000000
                                                                         480.000000
                 Credit History
         count
                       65.00000
                        0.60000
         mean
         std
                        0.49371
         min
                        0.00000
         25%
                        0.00000
         50%
                        1.00000
         75%
                        1.00000
         max
                        1.00000
```

```
In [1]: pip install scikit-learn
```

Requirement already satisfied: scikit-learn in c:\users\user\appdata\local\pr ograms\python\python311\lib\site-packages (1.2.2)
Requirement already satisfied: numpy>=1.17.3 in c:\users\user\appdata\local\pr rograms\python\python311\lib\site-packages (from scikit-learn) (1.24.3)
Requirement already satisfied: scipy>=1.3.2 in c:\users\user\appdata\local\pr ograms\python\python311\lib\site-packages (from scikit-learn) (1.10.1)
Requirement already satisfied: joblib>=1.1.1 in c:\users\user\appdata\local\pr rograms\python\python311\lib\site-packages (from scikit-learn) (1.3.0)
Requirement already satisfied: threadpoolctl>=2.0.0 in c:\users\user\appdata \local\programs\python\python311\lib\site-packages (from scikit-learn) (3.1.0)

Note: you may need to restart the kernel to use updated packages.

```
In [1]: import numpy as np  #Conduct a statistical test of the significance of the dimport scipy.stats as stats

# Define the two subsets of data
data_a = np.array([5849, 4583, 3000, 2583, 6000, 5417, 2333, 3036, 4006])
data_b = np.array([0, 128, 66, 120, 141, 267, 95, 158, 168])

# Conduct the independent t-test
t_statistic, p_value = stats.ttest_ind(data_a, data_b)

# Print the results
print("t-statistic:", t_statistic)
print("p-value:", p_value)
```

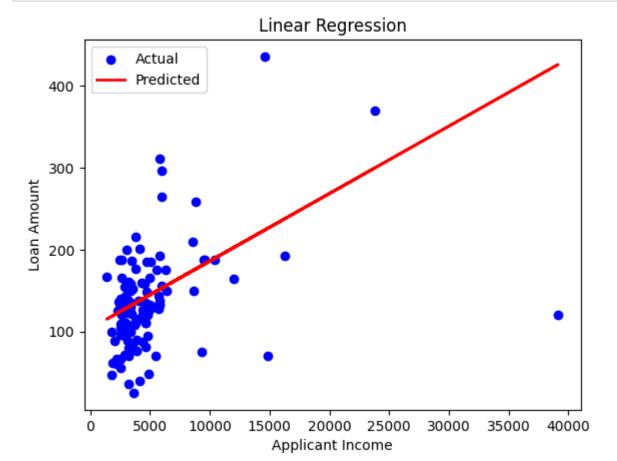
t-statistic: 8.289381635365402 p-value: 3.486682839706522e-07

```
In [9]: #Create one or more tables that group the data by a certain categorical variab
# Group the data by 'Gender' and calculate the mean and sum for selected column
grouped_gender = df.groupby('Gender').agg({'ApplicantIncome': 'mean', 'Coapplic
# Group the data by 'Education' and calculate the mean and sum for selected co
grouped_education = df.groupby('Education').agg({'ApplicantIncome': 'mean', 'Coapplic
# Print the grouped data for 'Gender'
print("Grouped Data by Gender:")
print(grouped_gender)
```

```
Grouped Data by Gender:
```

```
ApplicantIncome CoapplicantIncome LoanAmount Gender
Female 4593.954128 1138.504587 13810.0
Male 5455.895745 1764.442383 70155.0
```

```
In [7]: from sklearn.model selection import train test split
                                                                 #Implement a linear red
        from sklearn.linear model import LinearRegression
        # Remove rows with missing values
        df.dropna(subset=['LoanAmount'], inplace=True)
        # Split the data into training and testing sets
        X = df[['ApplicantIncome']]
        y = df['LoanAmount']
        X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, rando
        # Create and fit the linear regression model
        model = LinearRegression()
        model.fit(X_train, y_train)
        # Make predictions on the test set
        y_pred = model.predict(X_test)
        # Plot the actual vs. predicted values
        plt.scatter(X_test, y_test, color='blue', label='Actual')
        plt.plot(X test, y pred, color='red', linewidth=2, label='Predicted')
        plt.xlabel('Applicant Income')
        plt.ylabel('Loan Amount')
        plt.title('Linear Regression')
        plt.legend()
        plt.show()
        # Print the model coefficients and intercept
        print('Coefficients:', model.coef_)
        print('Intercept:', model.intercept_)
```



Coefficients: [0.00823739] Intercept: 103.70294803264798

In []: