

Case Study – 2

Python

Data Processing with Pandas

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1. Loading Data in Pandas DataFrame:

```
[21]: #Loading data into pandas
import pandas as pd
data = pd.read_csv('LoanData.csv') #Loading the data
print(data)
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	\
0	LP001002	Male	No	0	Graduate	No	
1	LP001003	Male	Yes	1	Graduate	No	
2	LP001005	Male	Yes	0	Graduate	Yes	
3	LP001006	Male	Yes	0	Not Graduate	No	
4	LP001008	Male	No	0	Graduate	No	
..	
609	LP002978	Female	No	0	Graduate	No	
610	LP002979	Male	Yes	3+	Graduate	No	
611	LP002983	Male	Yes	1	Graduate	No	
612	LP002984	Male	Yes	2	Graduate	No	
613	LP002990	Female	No	0	Graduate	Yes	

	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	\
0	5849	0.0	NaN	360.0	
1	4583	1508.0	128.0	360.0	
2	3000	0.0	66.0	360.0	
3	2583	2358.0	120.0	360.0	
4	6000	0.0	141.0	360.0	
..	
609	2900	0.0	71.0	360.0	
610	4106	0.0	40.0	180.0	
611	8072	240.0	253.0	360.0	
612	7583	0.0	187.0	360.0	
613	4583	0.0	133.0	360.0	

	Credit_History	Property_Area	Loan_Status
0	1.0	Urban	Y
1	1.0	Rural	N
2	1.0	Urban	Y
3	1.0	Urban	Y
4	1.0	Urban	Y
..
609	1.0	Rural	Y
610	1.0	Rural	Y
611	1.0	Urban	Y
612	1.0	Urban	Y
613	0.0	Semiurban	N

[614 rows x 13 columns]

2.Printing rows of the Data:

```
•[13]: #Printing rows of data & display values
display(data.head()) #first 5 rows
display(data.tail()) #last 5 rows
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Property
0	LP001002	Male	No	0	Graduate	No	5849	0.0	NaN	360.0	1.0	
1	LP001003	Male	Yes	1	Graduate	No	4583	1508.0	128.0	360.0	1.0	
2	LP001005	Male	Yes	0	Graduate	Yes	3000	0.0	66.0	360.0	1.0	
3	LP001006	Male	Yes	0	Not Graduate	No	2583	2358.0	120.0	360.0	1.0	
4	LP001008	Male	No	0	Graduate	No	6000	0.0	141.0	360.0	1.0	

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Property
609	LP002978	Female	No	0	Graduate	No	2900	0.0	71.0	360.0	1.0	
610	LP002979	Male	Yes	3+	Graduate	No	4106	0.0	40.0	180.0	1.0	
611	LP002983	Male	Yes	1	Graduate	No	8072	240.0	253.0	360.0	1.0	
612	LP002984	Male	Yes	2	Graduate	No	7583	0.0	187.0	360.0	1.0	
613	LP002990	Female	No	0	Graduate	Yes	4583	0.0	133.0	360.0	0.0	Set

3.Printing the column names of the DataFrame:

```
[20]: #Printing the column names of the DataFrame
display(list(data.columns))
```

```
['Loan_ID',
 'Gender',
 'Married',
 'Dependents',
 'Education',
 'Self_Employed',
 'ApplicantIncome',
 'CoapplicantIncome',
 'LoanAmount',
 'Loan_Amount_Term',
 'Credit_History',
 'Property_Area',
 'Loan_Status']
```

4.Summary of Data Frame:

```
[23]: #Summary of Data Frame
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 614 entries, 0 to 613
Data columns (total 13 columns):
#   Column              Non-Null Count  Dtype
---  -
0   Loan_ID              614 non-null    object
1   Gender               601 non-null    object
2   Married              611 non-null    object
3   Dependents           599 non-null    object
4   Education            614 non-null    object
5   Self_Employed        582 non-null    object
6   ApplicantIncome      614 non-null    int64
7   CoapplicantIncome    614 non-null    float64
8   LoanAmount           592 non-null    float64
9   Loan_Amount_Term     600 non-null    float64
10  Credit_History        564 non-null    float64
11  Property_Area        614 non-null    object
12  Loan_Status          614 non-null    object
dtypes: float64(4), int64(1), object(8)
memory usage: 62.5+ KB
```

5.Descriptive Statistical Measures of a DataFrame:

```
[24]: #Descriptive Statistical Measures of a DataFrame
data.describe()
```

[24]:	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

6. Missing Data Handling:

[25]: #Missing Data Handling
data.dropna()

[25]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Prope
1	LP001003	Male	Yes	1	Graduate	No	4583	1508.0	128.0	360.0	1.0	
2	LP001005	Male	Yes	0	Graduate	Yes	3000	0.0	66.0	360.0	1.0	
3	LP001006	Male	Yes	0	Not Graduate	No	2583	2358.0	120.0	360.0	1.0	
4	LP001008	Male	No	0	Graduate	No	6000	0.0	141.0	360.0	1.0	
5	LP001011	Male	Yes	2	Graduate	Yes	5417	4196.0	267.0	360.0	1.0	
...
609	LP002978	Female	No	0	Graduate	No	2900	0.0	71.0	360.0	1.0	
610	LP002979	Male	Yes	3+	Graduate	No	4106	0.0	40.0	180.0	1.0	
611	LP002983	Male	Yes	1	Graduate	No	8072	240.0	253.0	360.0	1.0	
612	LP002984	Male	Yes	2	Graduate	No	7583	0.0	187.0	360.0	1.0	
613	LP002990	Female	No	0	Graduate	Yes	4583	0.0	133.0	360.0	0.0	Se

480 rows × 13 columns

7. Sorting DataFrame values:

```
[58]: #Sorting DataFrame values
sorted_data = data.sort_values(by='ApplicantIncome')
print(sorted_data)
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	\
216	LP001722	Male	Yes	0	Graduate	No	
468	LP002502	Female	Yes	2	Not Graduate	NaN	
600	LP002949	Female	No	3+	Graduate	NaN	
500	LP002603	Female	No	0	Graduate	No	
188	LP001644	NaN	Yes	0	Graduate	Yes	
..	
185	LP001640	Male	Yes	0	Graduate	Yes	
155	LP001536	Male	Yes	3+	Graduate	No	
171	LP001585	NaN	Yes	3+	Graduate	No	
333	LP002101	Male	Yes	0	Graduate	NaN	
409	LP002317	Male	Yes	3+	Graduate	No	
	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	\		
216	150	1800.0	135.0	360.0			
468	210	2917.0	98.0	360.0			
600	416	41667.0	350.0	180.0			
500	645	3683.0	113.0	480.0			
188	674	5296.0	168.0	360.0			
..			
185	39147	4750.0	120.0	360.0			
155	39999	0.0	600.0	180.0			
171	51763	0.0	700.0	300.0			
333	63337	0.0	490.0	180.0			
409	81000	0.0	360.0	360.0			
	Credit_History	Property_Area	Loan_Status				
216	1.0	Rural	N				
468	1.0	Semiurban	Y				
600	NaN	Urban	N				
500	1.0	Rural	Y				
188	1.0	Rural	Y				

8.Merge Data Frames:

```
[59]: #Merge Data Frames
df1=pd.read_csv('LoanData.csv')
df2=pd.read_csv('LoanData.csv')

df=pd.merge(df1,df2)
print(df)
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	\
0	LP001002	Male	No	0	Graduate	No	
1	LP001003	Male	Yes	1	Graduate	No	
2	LP001005	Male	Yes	0	Graduate	Yes	
3	LP001006	Male	Yes	0	Not Graduate	No	
4	LP001008	Male	No	0	Graduate	No	
..	
609	LP002978	Female	No	0	Graduate	No	
610	LP002979	Male	Yes	3+	Graduate	No	
611	LP002983	Male	Yes	1	Graduate	No	
612	LP002984	Male	Yes	2	Graduate	No	
613	LP002990	Female	No	0	Graduate	Yes	

	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	\
0	5849	0.0	NaN	360.0	
1	4583	1508.0	128.0	360.0	
2	3000	0.0	66.0	360.0	
3	2583	2358.0	120.0	360.0	
4	6000	0.0	141.0	360.0	
..	
609	2900	0.0	71.0	360.0	
610	4106	0.0	40.0	180.0	
611	8072	240.0	253.0	360.0	
612	7583	0.0	187.0	360.0	
613	4583	0.0	133.0	360.0	

	Credit_History	Property_Area	Loan_Status
0	1.0	Urban	Y
1	1.0	Rural	N

9.Add new column to the Data Frame:

```
[60]: #Adding a new column to dataframe
data['newColumn']=10000
data.head()
```

```
[60]:
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Property
0	LP001002	Male	No	0	Graduate	No	5849	0.0	NaN	360.0	1.0	
1	LP001003	Male	Yes	1	Graduate	No	4583	1508.0	128.0	360.0	1.0	
2	LP001005	Male	Yes	0	Graduate	Yes	3000	0.0	66.0	360.0	1.0	
3	LP001006	Male	Yes	0	Not Graduate	No	2583	2358.0	120.0	360.0	1.0	
4	LP001008	Male	No	0	Graduate	No	6000	0.0	141.0	360.0	1.0	

10.Apply Function:

```
[61]: #Apply Function

def fun(value):
    if value>3000:
        return 'Yes'
    else:
        return 'No'

data['newColumn'] = data['ApplicantIncome'].apply(fun)
data.head()
```

```
[61]:
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Property
0	LP001002	Male	No	0	Graduate	No	5849	0.0	NaN	360.0	1.0	
1	LP001003	Male	Yes	1	Graduate	No	4583	1508.0	128.0	360.0	1.0	
2	LP001005	Male	Yes	0	Graduate	Yes	3000	0.0	66.0	360.0	1.0	
3	LP001006	Male	Yes	0	Not Graduate	No	2583	2358.0	120.0	360.0	1.0	
4	LP001008	Male	No	0	Graduate	No	6000	0.0	141.0	360.0	1.0	

11.By using the lambda operator:

```
[35]: #By using the lambda operator
data['outputColumn'] = data['LoanAmount'].apply(lambda x:x/10)
data['apColumn'] = data['ApplicantIncome'].apply(lambda x:x/10)
data.head()
```

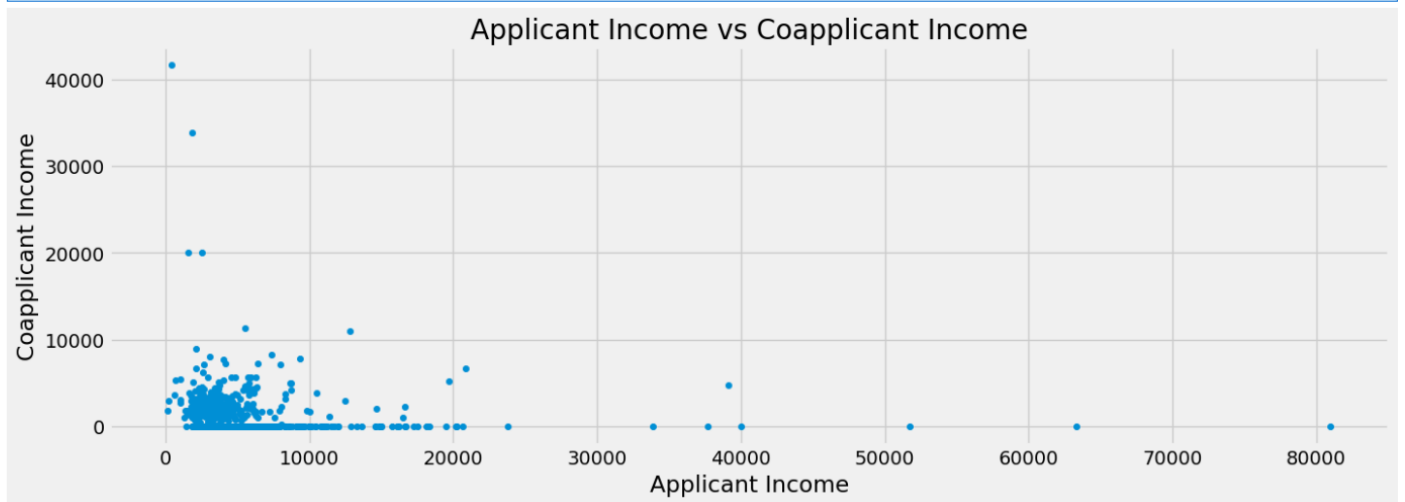
```
[35]:
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Property
0	LP001002	Male	No	0	Graduate	No	5849	0.0	NaN	360.0	1.0	
1	LP001003	Male	Yes	1	Graduate	No	4583	1508.0	128.0	360.0	1.0	
2	LP001005	Male	Yes	0	Graduate	Yes	3000	0.0	66.0	360.0	1.0	
3	LP001006	Male	Yes	0	Not Graduate	No	2583	2358.0	120.0	360.0	1.0	
4	LP001008	Male	No	0	Graduate	No	6000	0.0	141.0	360.0	1.0	

12.Visualizing DataFrame:

```
[55]: #Visualizing DataFrame
import matplotlib.pyplot as plt
data.plot( x='ApplicantIncome',y='CoapplicantIncome',kind='scatter')

plt.title("Applicant Income vs Coapplicant Income")
plt.xlabel("Applicant Income")
plt.ylabel("Coapplicant Income")
plt.show()
```



Submitted by:
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