

Python Case Study

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Problem Statement :

Automate the loan eligibility process (real-time) based on customer detail provided while filling

the online application form. These details are Gender, Marital Status, Education, Number of Dependents, Income, Loan Amount, Credit History, and others.


The major aim of this notebook is to predict which of the customers will have their loan approved.

● Loading Data in Pandas DataFrame

```
import pandas as pd
```

```
df = pd.read_csv("/content/sample_data/LoanData.csv")
```


```
df.head()
```



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

● Printing rows of the Data

```
df.columns
```



```
Index(['Loan_ID', 'Gender', 'Married', 'Dependents', 'Education',  
      'Self_Employed', 'ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',  
      'Loan_Amount_Term', 'Credit_History', 'Property_Area', 'Loan_Status'],  
      dtype='object')
```

● Summary of Data Frame

```
df.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

```


- Descriptive Statistical Measures of a DataFrame

df.describe()

	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History
count	614.000000	614.000000	592.000000	600.000000	564.000000
mean	5403.459283	1621.245798	146.412162	342.000000	0.842199
std	6109.041673	2926.248369	85.587325	65.12041	0.364878
min	150.000000	0.000000	9.000000	12.000000	0.000000
25%	2877.500000	0.000000	100.000000	360.000000	1.000000
50%	3812.500000	1188.500000	128.000000	360.000000	1.000000
75%	5795.000000	2297.250000	168.000000	360.000000	1.000000
max	81000.000000	41667.000000	700.000000	480.000000	1.000000

- Missing Data Handling

df.isnull().sum()




	8
Loan_ID	0
Gender	13
Married	3
Dependents	15
Education	0
Self_Employed	32
ApplicantIncome	0
CoapplicantIncome	0
LoanAmount	22
Loan_Amount_Term	14
Credit_History	50
Property_Area	0
Loan_Status	0

dtype: int64

- Sorting DataFrame values

`df.sort_values(by='ApplicantIncome', ascending=False)`



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

614 rows × 13 columns

- Merge Data Frames

```
df1 = pd.read_csv("/content/sample_data/LoanData.csv")
```

```
df2 = pd.read_csv("/content/sample_data/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
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]

- Apply Function

```
df['LoanAmount_k'] = df['LoanAmount'].apply(lambda x: x / 1000)
```

```
df.head()
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Property_Area	Loan_Status	LoanAmount_k
0	LP001002	Male	No	0	Graduate	No	5849	0.0	NaN	360.0	1.0	Urban	Y	NaN
1	LP001003	Male	Yes	1	Graduate	No	4583	1508.0	128.0	360.0	1.0	Rural	N	0.128
2	LP001005	Male	Yes	0	Graduate	Yes	3000	0.0	66.0	360.0	1.0	Urban	Y	0.066
3	LP001006	Male	Yes	0	Not Graduate	No	2583	2358.0	120.0	360.0	1.0	Urban	Y	0.120
4	LP001008	Male	No	0	Graduate	No	6000	0.0	141.0	360.0	1.0	Urban	Y	0.141

- By using the lambda operator

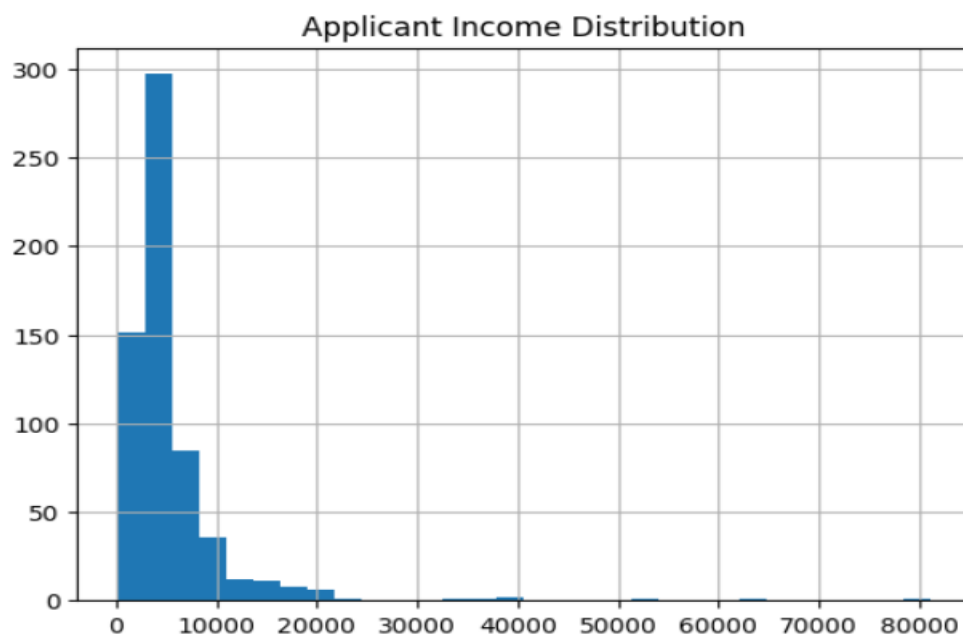
```
df['IncomeLevel'] = df['ApplicantIncome'].apply(lambda x: 'High' if x > 5000 else 'Low')
```

```
df.head()
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Property_Area	Loan_Status	LoanAmount_k	IncomeLevel
0	LP001002	Male	No	0	Graduate	No	5849	0.0	NaN	360.0	1.0	Urban	Y	NaN	High
1	LP001003	Male	Yes	1	Graduate	No	4583	1508.0	128.0	360.0	1.0	Rural	N	0.128	Low
2	LP001005	Male	Yes	0	Graduate	Yes	3000	0.0	66.0	360.0	1.0	Urban	Y	0.066	Low
3	LP001006	Male	Yes	0	Not Graduate	No	2583	2358.0	120.0	360.0	1.0	Urban	Y	0.120	Low
4	LP001008	Male	No	0	Graduate	No	6000	0.0	141.0	360.0	1.0	Urban	Y	0.141	High

- Visualizing DataFrame

```
df['ApplicantIncome'].hist(bins=30)  
plt.title("Applicant Income Distribution")  
plt.show()
```



```
import matplotlib.pyplot as plt  
import seaborn as sns  
sns.countplot(x='Loan_Status', data=df)  
plt.title('Loan Approval Count')  
plt.show()
```

