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In [1]: import pandas as pd
import numpy as np

In [2]: df = pd.read_csv("Dataset.csv")

In [3]: df.head()

Out[3]:
   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

In [4]: df.shape

Out[4]:
(614, 13)

In [5]: 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

In [6]: df.describe()

   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 [7]: df['Gender'].value_counts()

Male      489
Female    112
Name: Gender, dtype: int64

In [8]: df['Married'].value_counts()

Yes       398
No        213
Name: Married, dtype: int64

In [9]: df['Dependents'].value_counts()

0         345
1         102
2         101
3+         51
Name: Dependents, dtype: int64

In [10]: df['Education'].value_counts()

Graduate      480
Not Graduate   134
Name: Education, dtype: int64

In [11]: df['Self_Employed'].value_counts()

No        500
Yes        82
Name: Self_Employed, dtype: int64

In [12]: 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')

In [13]: df.isnull().sum()

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

In [14]: df1=df.copy()

In [15]: df1= df1.drop(columns=['Loan_ID'])

In [16]: df1.shape

Out[16]:
(614, 12)

In [17]: df1['LoanAmount']=df1['LoanAmount'].fillna(df1['LoanAmount'].mean())

In [18]: df1['Loan_Amount_Term']=df1['Loan_Amount_Term'].fillna(df1['Loan_Amount_Term'].mean())

In [19]: df1['Credit_History']=df1['Credit_History'].fillna(df1['Credit_History'].mean())

In [20]: df1['Gender']=df1['Gender'].fillna(df1['Gender'].mode()[0])

In [21]: df1['Married']=df1['Married'].fillna(df1['Married'].mode()[0])

In [22]: df1['Dependents']=df1['Dependents'].fillna(df1['Dependents'].mode()[0])

In [23]: df1['Self_Employed']=df1['Self_Employed'].fillna(df1['Self_Employed'].mode()[0])

In [24]: df1.isnull().sum()

Gender      0
Married     0
Dependents  0
Education   0
Self_Employed 0
ApplicantIncome 0
CoapplicantIncome 0
LoanAmount   0
Loan_Amount_Term 0
Credit_History 0
Property_Area 0
Loan_Status  0
dtype: int64

In [25]: df1.dtypes

Gender      object
Married     object
Dependents  object
Education   object
Self_Employed object
ApplicantIncome int64
CoapplicantIncome float64
LoanAmount   float64
Loan_Amount_Term float64
Credit_History float64
Property_Area object
Loan_Status  object
dtype: object

In [26]: from sklearn import preprocessing
label_encoder = preprocessing.LabelEncoder()

In [27]: df1['Gender']= label_encoder.fit_transform(df1['Gender'])

In [28]: df1['Married']= label_encoder.fit_transform(df1['Married'])

In [29]: df1['Education']= label_encoder.fit_transform(df1['Education'])

In [30]: df1['Self_Employed']= label_encoder.fit_transform(df1['Self_Employed'])

In [31]: df1['Property_Area']= label_encoder.fit_transform(df1['Property_Area'])

In [32]: df1['Loan_Status']= label_encoder.fit_transform(df1['Loan_Status'])

In [33]: df1.dtypes

Gender      int32
Married     int32
Dependents  object
Education   int32
Self_Employed int32
ApplicantIncome int64
CoapplicantIncome float64
LoanAmount   float64
Loan_Amount_Term float64
Credit_History float64
Property_Area int32
Loan_Status  int32
dtype: object

In [34]: df1.head(10)

Out[34]:
   Gender  Married  Dependents  Education  Self_Employed  ApplicantIncome  CoapplicantIncome  LoanAmount  Loan_Amount_Term  Credit_History  Property_Area  Loan_Status
0      1         0           0          0           0           5849             0.0          146.412162           360.0           1.0           2           1
1      1         1           1          0           0           4583            1508.0          128.000000           360.0           1.0           0           0
2      1         1           0          0           1           3000             0.0          66.000000           360.0           1.0           2           1
3      1         1           0          1           0           2583            2358.0          120.000000           360.0           1.0           2           1
4      1         0           0          0           0           6000             0.0          141.000000           360.0           1.0           2           1
5      1         1           2          0           1           5417            4196.0          267.000000           360.0           1.0           2           1
6      1         1           0          1           0           2333            1516.0          95.000000           360.0           1.0           2           1
7      1         1           3+          0           0           3036            2504.0          158.000000           360.0           0.0           1           0
8      1         1           2          0           0           4006            1526.0          168.000000           360.0           1.0           2           1
9      1         1           1          0           0          12841            10968.0          349.000000           360.0           1.0           1           0
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