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```
1 import numpy as np
2 import pandas as pd

1 df = pd.read_csv("/content/train_ctrUa4K.csv")
2 df.head()
```

↗

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome
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

```
1 df.tail()
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome
609	LP002978	Female	No	0	Graduate	No	290
610	LP002979	Male	Yes	3+	Graduate	No	410
611	LP002983	Male	Yes	1	Graduate	No	807
612	LP002984	Male	Yes	2	Graduate	No	758
613	LP002990	Female	No	0	Graduate	Yes	458

```
1 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
```

```
1 df["Gender"].unique()  
  
array(['Male', 'Female', nan], dtype=object)
```

```
1 df["Gender"] = df["Gender"].replace(["Female", "Male"], [1, 0])  
2 df["Gender"] = df["Gender"].replace([np.nan],0)  
3 df["Gender"].unique()  
  
array([0., 1.])
```

```
1 df["Gender"].isnull().sum()  
  
0
```

```
1 df["Married"].unique()  
  
array(['No', 'Yes', nan], dtype=object)
```

```
1 df["Married"] = df["Married"].replace(["No", "Yes"], [0, 1])  
2 df["Married"] = df["Married"].replace([np.nan],0)  
3 df["Married"].unique()  
  
array([0., 1.])
```

```
1 df["Married"].isnull().sum()  
  
0
```

```
1 df["Dependents"].unique()  
  
array(['0', '1', '2', '3+', nan], dtype=object)
```

```
1 df["Dependents"] = df["Dependents"].replace([np.nan, '3+', "1", "2"],[0, 3, 1, 2])  
2 df["Dependents"].unique()  
  
array([0, 1, 2, 3])
```

```
1 df["Dependents"].isnull().sum()  
  
0
```

```
1 df["Self_Employed"].unique()  
  
array(['No', 'Yes', nan], dtype=object)
```

```
1 df["Self_Employed"] = df["Self_Employed"].replace(["No", "Yes"], [0, 1])
```

```
1 df["Self_Employed"] = df["Self_Employed"].replace([no, yes], [0, 1])
2 df["Self_Employed"] = df["Self_Employed"].replace([np.nan], 0)
3 df["Self_Employed"].unique()
```

```
array([0., 1.])
```

```
1 df["Self_Employed"].isnull().sum()
```

```
0
```

```
1 df["LoanAmount"] = df["LoanAmount"].fillna(method = "ffill")
2 df["LoanAmount"] = df["LoanAmount"].replace([np.nan], 0)
3 df["LoanAmount"].isnull().sum()
```

```
0
```

```
1 df["LoanAmount"].unique()
```

```
array([ 0., 128., 66., 120., 141., 267., 95., 158., 168., 349., 70.,
       109., 200., 114., 17., 125., 100., 76., 133., 115., 104., 315.,
       116., 112., 151., 191., 122., 110., 35., 201., 74., 106., 320.,
       144., 184., 80., 47., 75., 134., 96., 88., 44., 286., 97.,
       135., 180., 99., 165., 258., 126., 312., 136., 172., 81., 187.,
       113., 176., 130., 111., 167., 265., 50., 210., 175., 131., 188.,
        25., 137., 160., 225., 216., 94., 139., 152., 118., 185., 154.,
        85., 259., 194., 93., 370., 182., 650., 102., 290., 84., 242.,
       129., 30., 244., 600., 255., 98., 275., 121., 63., 700., 87.,
       101., 495., 67., 73., 260., 108., 58., 48., 164., 170., 83.,
        90., 166., 124., 55., 59., 127., 214., 240., 72., 60., 138.,
        42., 280., 140., 155., 123., 279., 192., 304., 330., 150., 207.,
       436., 78., 54., 89., 143., 105., 132., 480., 56., 159., 300.,
       376., 117., 71., 490., 173., 46., 228., 308., 236., 570., 380.,
       296., 156., 103., 45., 65., 53., 360., 62., 218., 178., 239.,
       405., 148., 190., 149., 153., 162., 230., 86., 234., 246., 500.,
       186., 119., 107., 209., 208., 243., 40., 250., 311., 400., 161.,
       196., 324., 157., 145., 181., 26., 211., 9., 205., 36., 61.,
       146., 292., 142., 350., 496., 253.]])
```

```
1 df["Loan_Amount_Term"] = df["Loan_Amount_Term"].fillna(method = "ffill")
2 df["Loan_Amount_Term"].unique()
```

```
array([360., 120., 240., 180., 60., 300., 480., 36., 84., 12.]])
```

```
1 df["Loan_Amount_Term"].isnull().sum()
```

```
0
```

```
1 df["Credit_History"].unique()
```

```
array([ 1., 0., nan])
```

```
1 df["Credit_History"] = df["Credit_History"].replace([np.nan], 0)
2 df["Credit_History"].unique()
```

```
array([1., 0.])
```

```
1 df["Credit_History"].isnull().sum()
```

```
0
```

```
1 df.isnull().sum()
```

```
Loan_ID          0
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
```

```
1 df["Education"].unique()
```

```
2 df["Education"] = df["Education"].replace(["Graduate", "Not Graduate"], [1, 0])
```

```
3 df["Education"].unique()
```

```
array([1, 0])
```

```
1 df["Property_Area"].unique()
```

```
array(['Urban', 'Rural', 'Semiurban'], dtype=object)
```

```
1 df["Property_Area"] = df["Property_Area"].replace(["Urban", "Rural", "Semiurban"], [1,
```

```
2 df["Property_Area"].unique()
```

```
array([1, 2, 3])
```

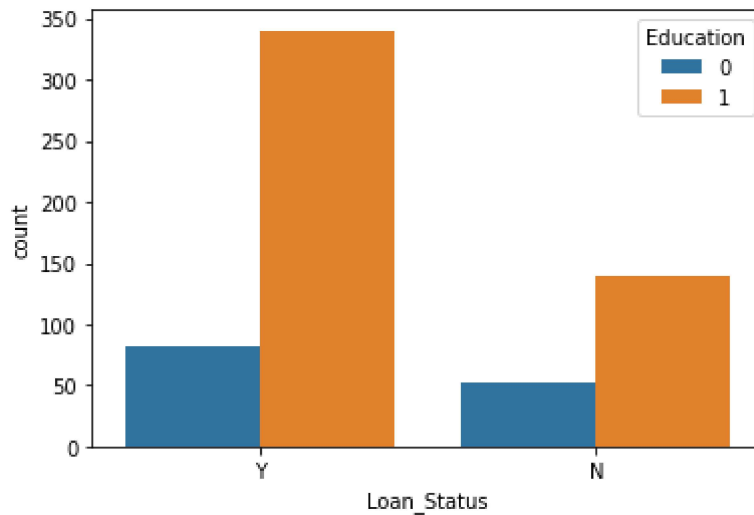
```
1 import matplotlib.pyplot as plt
```

```
2 import seaborn as sb
```

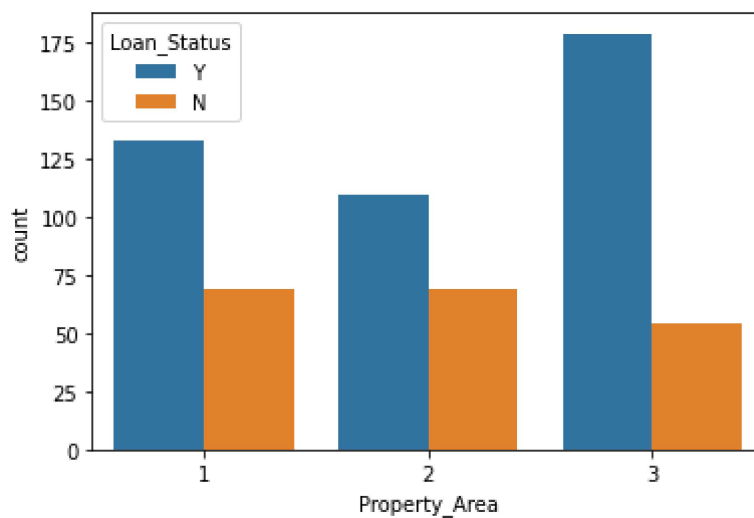
```
1 sb.countplot(x = "Education", data = df);
```



```
1 sb.countplot(x = "Loan_Status", hue = "Education", data = df);
```



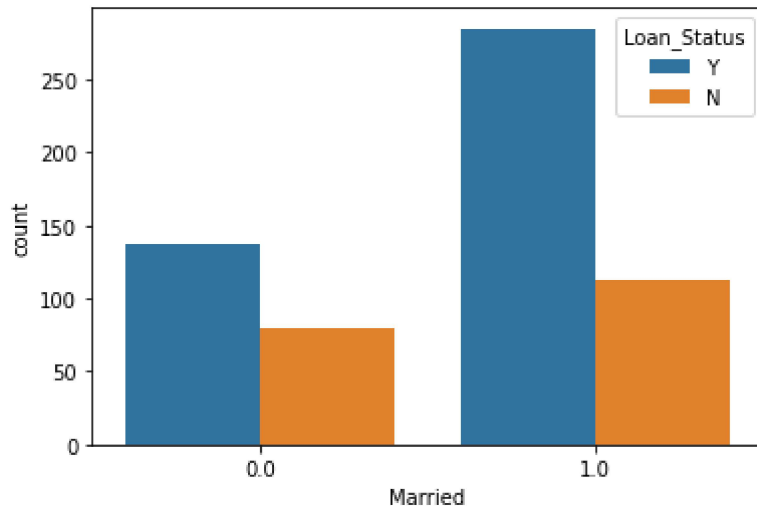
```
1 sb.countplot(x = "Property_Area", hue = "Loan_Status", data = df);
```



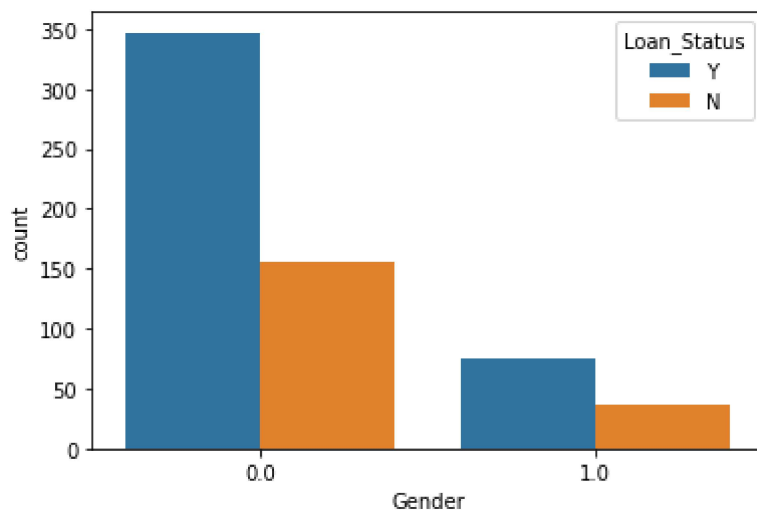
```
1 sb.countplot(x = "Self_Employed", hue = "Loan_Status", data = df);
```



```
1 sb.countplot(x = "Married", hue = "Loan_Status", data = df);
```



```
1 sb.countplot(x = "Gender", hue = "Loan_Status", data = df);
```



```
1 x = df[["Gender", "Married", "Dependents", "Education", "Self_Employed", "ApplicantInco
2 y = df["Loan_Status"]
```

```
1 from sklearn.model_selection import train_test_split
2 xtrain, xtest, ytrain, ytest = train_test_split(x, y, test_size = 0.3, random_state = 3
3 print(xtrain.shape)
4 print(ytrain.shape)
5 print(xtest.shape)
6 print(ytest.shape)
```

```
(429, 11)
(429,)
(185, 11)
(185,)
```

```
1 from sklearn.tree import DecisionTreeClassifier
2 model = DecisionTreeClassifier(max_depth = 4)
```

```
1 model.fit(xtrain, ytrain)
2 ypred = model.predict(xtest)
```

```
1 from sklearn.metrics import accuracy_score
2 accuracy = accuracy_score(ypred, ytest)
3 print(accuracy*100)
```

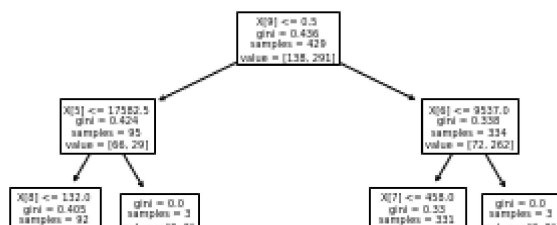
76.21621621621621

```
1 from sklearn.metrics import confusion_matrix
2 matrix = confusion_matrix(ypred, ytest)
3 print(matrix)
```

```
[[ 28  18]
 [ 26 113]]
```

```
1 from sklearn import tree
2 tree.plot_tree(model)
3 tree.export_graphviz(model)
```

```
'digraph Tree {\nnode [shape=box] ;\n0 [label="X[9] <= 0.5\\ngini = 0.436\\nsamples = 429\\nvalue = [138, 291]" ] ;\n1 [label="X[5] <= 17582.5\\ngini = 0.424\\nsamples = 95\\nvalue = [66, 29]" ] ;\n0 -> 1 [labeldistance=2.5, labelangle=45, headlabel="True" ] ;\n2 [label="X[8] <= 132.0\\ngini = 0.405\\nsamples = 92\\nvalue = [66, 26]" ] ;\n1 -> 2 ;\n3 [label="gini = 0.0\\nsamples = 2\\nvalue = [0, 2]" ] ;\n2 -> 3 ;\n4 [label="X[5] <= 2554.5\\ngini = 0.391\\nsamples = 90\\nvalue = [66, 24]" ] ;\n2 -> 4 ;\n5 [label="gini = 0.498\\nsamples = 17\\nvalue = [9, 8]" ] ;\n4 -> 5 ;\n6 [label="gini = 0.342\\nsamples = 73\\nvalue = [57, 16]" ] ;\n4 -> 6 ;\n7 [label="gini = 0.0\\nsamples = 3\\nvalue = [0, 3]" ] ;\n6 -> 7 ;\n8 [label="X[6] <= 9537.0\\ngini = 0.338\\nsamples = 334\\nvalue = [72, 262]" ] ;\n0 -> 8 [labeldistance=2.5, labelangle=-45, headlabel="False" ] ;\n9 [label="X[7] <= 458.0\\ngini = 0.33\\nsamples = 331\\nvalue = [69, 262]" ] ;\n8 -> 9 ;\n10 [label="X[10] <= 2.5\\ngini = 0.317\\nsamples = 324..." ] ;\n9 -> 10 ;\n}
```



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