

OUTPUT SCREEN

The screenshot shows a Jupyter Notebook interface with two code cells. The first cell, labeled [6], contains the code `data.head()` and has a runtime time of 0.8s. The second cell, labeled [7], contains the code `data.tail()` and has a runtime time of 0.1s. Both cells display a table of loan data with 13 columns: Loan_ID, Gender, Married, Dependents, Education, Self_Employed, ApplicantIncome, CoapplicantIncome, LoanAmount, Loan_Amount_Term, Credit_History, and Property_Area. The first cell shows the first 5 rows (Loan_IDs LP001002 to LP001008), and the second cell shows the last 5 rows (Loan_IDs LP002978 to LP002990).

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

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

The screenshot shows a Jupyter Notebook interface with two code cells. The first cell, labeled [9], contains the code `data.shape` and has a runtime time of 0.2s. The second cell, labeled [10], contains the code `data.info()` and has a runtime time of 0.9s. The first cell displays the output `(614, 13)`. The second cell displays the output of the `data.info()` method, showing the data type of the DataFrame, the range of indices, the number of columns, and the data types of each column.

```
<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
```

```
data.isnull().sum()
```

[11] ✓ 0.1s Python

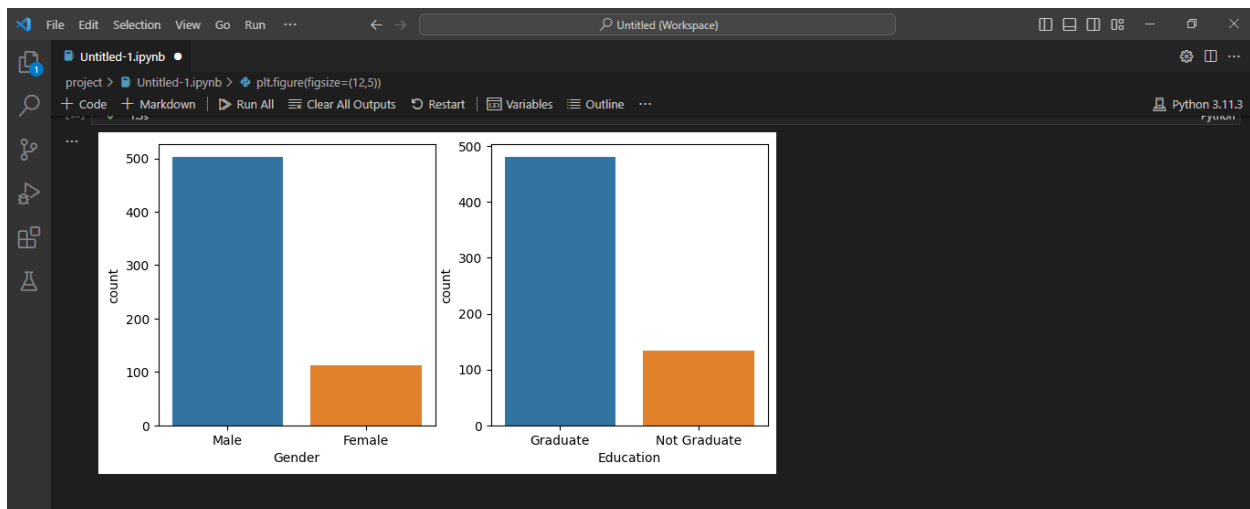
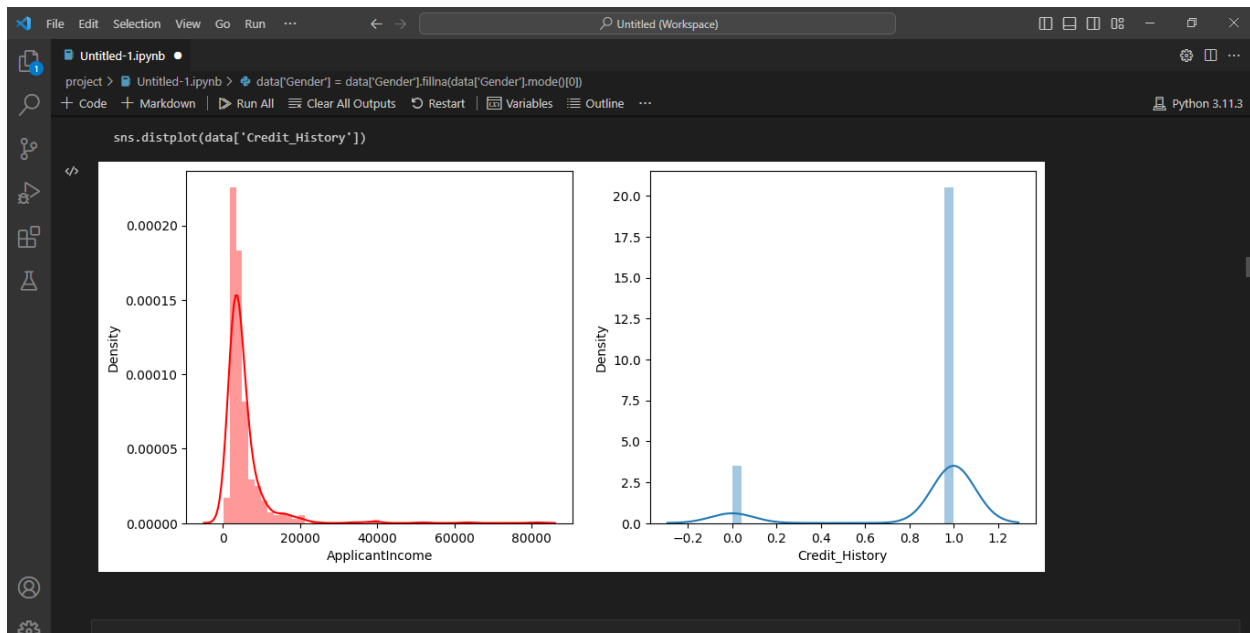
```
... Loan_ID      0
    Gender      13
    Married      3
    Dependents   15
    Education     0
    Self_Employed 32
    ApplicantIncome 0
    CoapplicantIncome 0
    LoanAmount    22
    Loan_Amount_Term 14
    Credit_History 50
    Property_Area  0
    Loan_Status    0
    dtype: int64
```

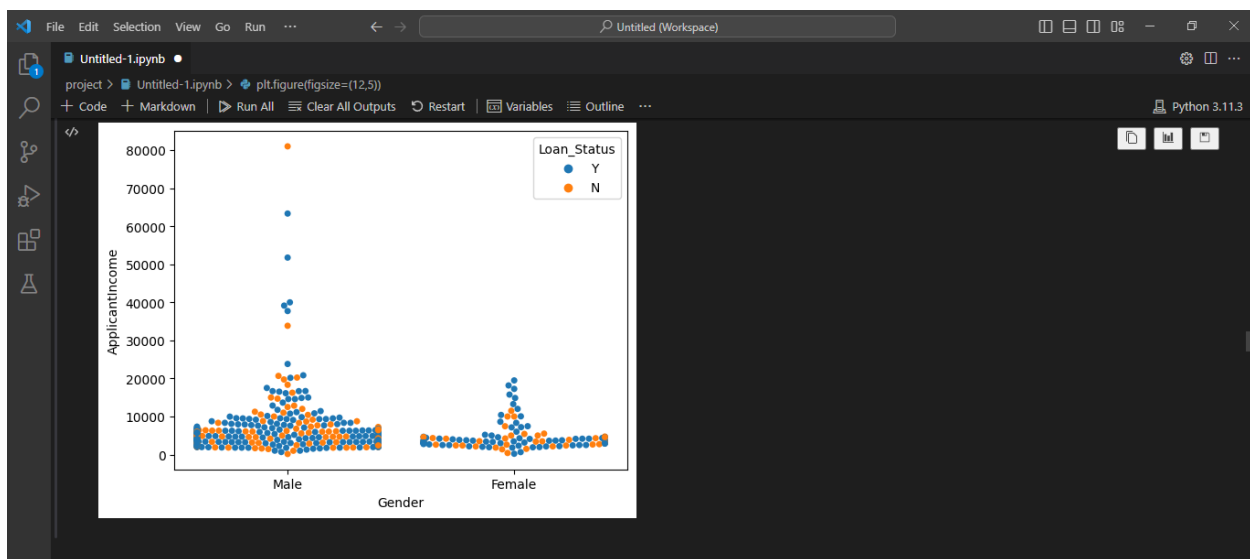
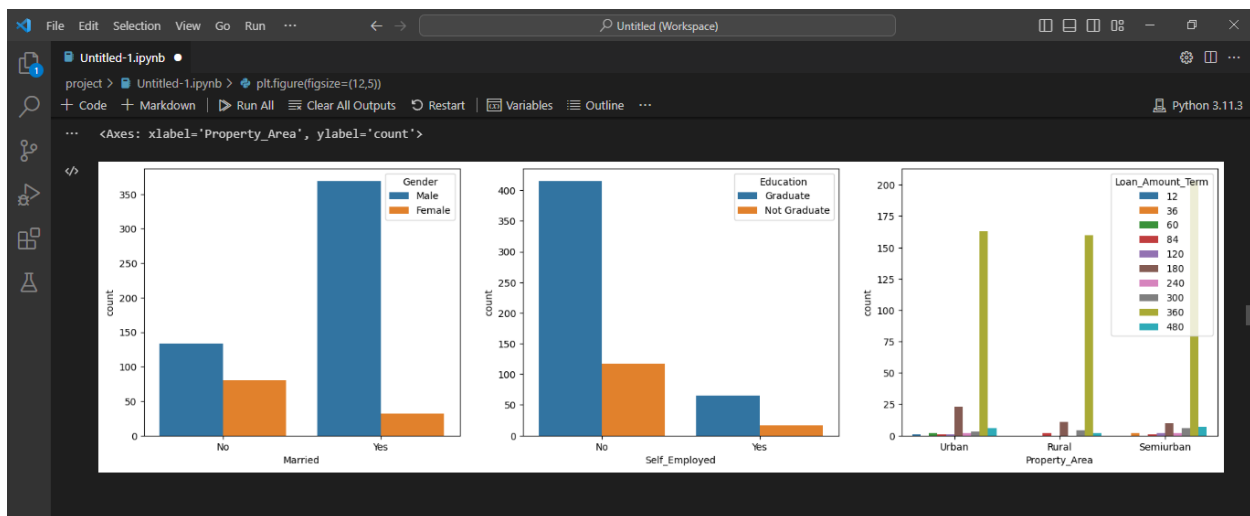
```
data["Gender"] = data["Gender"].fillna(data["Gender"].mode[0])
```

```
data.describe()
```

[16] ✓ 0.7s Python

	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History
count	614.000000	614.000000	614.000000	614.000000	614.000000
mean	5403.459283	1621.24430	145.465798	342.410423	0.855049
std	6109.041673	2926.24876	84.180967	64.428629	0.352339
min	150.000000	0.000000	9.000000	12.000000	0.000000
25%	2877.500000	0.000000	100.250000	360.000000	1.000000
50%	3812.500000	1188.500000	125.000000	360.000000	1.000000
75%	5795.000000	2297.250000	164.750000	360.000000	1.000000
max	81000.000000	41667.000000	700.000000	480.000000	1.000000





Untitled-1.ipynb •

project > Untitled-1.ipynb > plt.figure(figsize=(12,5))

+ Code + Markdown ▶ Run All ⌵ Clear All Outputs ↺ Restart 📄 Variables 📄 Outline ...

Python 3.11.3

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Property_Area
0	LP001002	1	0	0	0	0	5849	0	120	360	1	2
1	LP001003	1	1	1	0	0	4583	1508	128	360	1	0
2	LP001005	1	1	0	0	1	3000	0	66	360	1	2
3	LP001006	1	1	0	1	0	2583	2358	120	360	1	2
4	LP001008	1	0	0	0	0	6000	0	141	360	1	2


```
File Edit Selection View Go Run ... Untitled (Workspace) Python 3.11.3
project > Untitled-1.ipynb > compareModel(X_train,X_test,Y_train,Y_test)
+ Code + Markdown | Run All | Clear All Outputs | Restart | Variables | Outline ...
[47] ✓ 2.1s
...
***Decision TreeClassifier***
Confusion matrix
[[ 31  41]
 [ 31 100]]
Classification report
precision    recall  f1-score   support

     0       0.50      0.43      0.46         72
     1       0.71      0.76      0.74        131

 accuracy      0.60      0.60      0.65        203
 macro avg     0.60      0.60      0.60        203
 weighted avg   0.64      0.65      0.64        203

-----
***RandomForestClassifier***
Confusion matrix
[[ 39  33]
 [  8 123]]
Classification report
precision    recall  f1-score   support

     0       0.83      0.54      0.66         72
     1       0.79      0.94      0.86        131

 accuracy      0.80      0.80      0.80        203
 macro avg     0.81      0.74      0.76        203
 weighted avg   0.80      0.80      0.79        203
```

```
File Edit Selection View Go Run ... Untitled (Workspace) Python 3.11.3
project > Untitled-1.ipynb > compareModel(X_train,X_test,Y_train,Y_test)
+ Code + Markdown | Run All | Clear All Outputs | Restart | Variables | Outline ...
Cell 47 of 55
-----
***KNeighborsClassifier***
Confusion matrix
[[  8  64]
 [ 24 107]]
Classification report
precision    recall  f1-score   support

     0       0.25      0.11      0.15         72
     1       0.63      0.82      0.71        131

 accuracy      0.44      0.46      0.57        203
 macro avg     0.44      0.46      0.43        203
 weighted avg   0.49      0.57      0.51        203

-----
***GradientBoostingClassifier***
Confusion matrix
[[ 32  40]
 [ 10 121]]
Classification report
precision    recall  f1-score   support

     0       0.76      0.44      0.56         72
     1       0.75      0.92      0.83        131

 accuracy      0.76      0.68      0.75        203
 macro avg     0.76      0.68      0.70        203
 weighted avg   0.76      0.75      0.73        203
```

```
File Edit Selection View Go Run ... Untitled (Workspace)
Untitled-1.ipynb
project > Untitled-1.ipynb > def compareModel(X_train,X_test,Y_train,Y_test):
+ Code + Markdown | Run All | Clear All Outputs | Restart | Variables | Outline ... Python 3.11.3

ypred = classifier.predict(X_test)
print(accuracy_score(y_pred,Y_test))
print("ANN Model")
print("Confusion_Matrix")
print(confusion_matrix(Y_test,y_pred))
print("Classification_Report")
print(classification_report(Y_test,y_pred))

[48] ✓ 0.4s Python

... 7/7 [=====] - 0s 5ms/step
0.625615763546798
ANN Model
Confusion_Matrix
[[ 3  69]
 [ 7 124]]
Classification_Report
precision    recall  f1-score   support

      0       0.30      0.04      0.07        72
      1       0.64      0.95      0.77       131

 accuracy          0.47      0.49      0.42       203
 macro avg          0.47      0.49      0.42       203
 weighted avg        0.52      0.63      0.52       203

from sklearn.model_selection import cross_val_score
```

```
f1_score(y_pred,Y_test,average='weighted')
[51] ✓ 0.1s Python
... 0.7313305535901553

cv = cross_val_score(rf,x,y,cv=5)
[52] ✓ 4.4s Python

np.mean(cv)
[53] ✓ 0.2s Python
... 0.7833799813407971
```