

```
# importing the required libraries
import pandas as pd
import numpy as np
import sklearn
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score
import matplotlib.pyplot as plt
%matplotlib inline

#CHECKING VERSION OF SKLEARN
print(sklearn.__version__)

0.22.2.post1

#loading the preprocessed dataset
data=pd.read_csv("/content/Loan_Prediction_New_Data.csv")

data.head()
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome
0	LP001002	0.0	0.0	0.000000	1.0	0.0	0.07048
1	LP001003	0.0	1.0	0.333333	1.0	0.0	0.0548
2	LP001005	0.0	1.0	0.000000	1.0	1.0	0.0352
3	LP001006	0.0	1.0	0.000000	0.0	0.0	0.0300
4	LP001008	0.0	0.0	0.000000	1.0	0.0	0.0723

```
#checking missing values
data.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
```

```
#checking data types
data.dtypes
```

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

```
#dropint the loan id since it is just a unique value
data=data.drop("Loan_ID",axis=1)
```

```
data.head()
```

	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome
0	0.0	0.0	0.000000	1.0	0.0	0.070489	0.000000
1	0.0	1.0	0.333333	1.0	0.0	0.054830	0.000000
2	0.0	1.0	0.000000	1.0	1.0	0.035250	0.000000
3	0.0	1.0	0.000000	0.0	0.0	0.030093	0.000000
4	0.0	0.0	0.000000	1.0	0.0	0.072356	0.000000

```
data.shape
```

```
(614, 12)
```

```
#SEPERATE THE DATA INTO DEPENDENT AND INDEPENDENT
x=data.drop("Loan_Status",axis=1)
```

```
#independent variable
y
```

	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome
0	0.0	0.0	0.000000	1.0	0.0	0.070489	0
1	0.0	1.0	0.333333	1.0	0.0	0.054830	0
2	0.0	1.0	0.000000	1.0	1.0	0.035250	0
3	0.0	1.0	0.000000	0.0	0.0	0.030093	0
4	0.0	0.0	0.000000	1.0	0.0	0.072356	0
...
609	1.0	0.0	0.000000	1.0	0.0	0.034014	0
610	0.0	1.0	1.000000	1.0	0.0	0.048930	0
611	0.0	1.0	0.333333	1.0	0.0	0.097984	0
612	0.0	1.0	0.666667	1.0	0.0	0.091936	0

```
y=data["Loan_Status"]
```

```
614 rows x 11 columns
```

```
#dependent variable(target variable)
```

```
y
```

```
0      1.0
1      0.0
2      1.0
3      1.0
4      1.0
```

```
...
609    1.0
610    1.0
611    1.0
612    1.0
613    0.0
```

```
Name: Loan_Status, Length: 614, dtype: float64
```

```
#shape of independent and dependent variable
```

```
x.shape,y.shape
```

```
((614, 11), (614,))
```

CREATING TEST AND TRAINING DATA

```
xtrain,xtest,ytrain,ytest=train_test_split(x,y,stratify=data["Loan_Status"],random_state=10,t
```

```
#shape of training and test set
```

```
(xtrain.shape,ytrain.shape),(xtest.shape,ytest.shape)
```

(((491, 11), (491,)), ((123, 11), (123,)))

xtrain

	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome
164	0.0	1.0	0.000000	1.0	0.0	0.113457	0
171	0.0	1.0	1.000000	1.0	0.0	0.638380	0
546	0.0	0.0	0.000000	0.0	0.0	0.039678	0
226	0.0	1.0	0.000000	0.0	1.0	0.056710	0
176	0.0	1.0	0.666667	1.0	0.0	0.023438	0
...
560	0.0	1.0	0.666667	0.0	0.0	0.043599	0
503	0.0	1.0	0.333333	0.0	0.0	0.048237	0
343	0.0	1.0	1.000000	0.0	0.0	0.037390	0
148	1.0	0.0	0.000000	1.0	0.0	0.121831	0
303	0.0	1.0	0.333333	1.0	0.0	0.018244	0

491 rows × 11 columns

xtest

	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	Loan_Status
--	--------	---------	------------	-----------	---------------	-----------------	-------------------	-------------

	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	Loan_Status
164	1.0							
171	1.0							
546	0.0							
226	0.0							
176	1.0							
...								
560	1.0							
503	0.0							
343	1.0							
148	0.0							
303	1.0							
Name: Loan_Status, Length: 491, dtype: float64								
369	0.0	1.0	0.000000	1.0	0.0	0.242177	0.0	0

	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	Loan_Status
507	0.0							
493	1.0							
434	1.0							
125	1.0							
294	1.0							
...								
82	0.0							
295	1.0							
369	0.0							
450	0.0							
363	1.0							
Name: Loan_Status, Length: 123, dtype: float64								

DEFINING ARCHITECTURE OF THE MODEL

```
#importing keras
```

```
import keras
```

```
#checking the version of the keras
```

```
print(keras.__version__)
```

```
2.5.0
```

```
#importing tensorflow
```

```
import tensorflow as tf
```

```
#checking the version of the tensorflow
```

```
print(tf.__version__)
```

```
2.5.0
```

```
#importing sequential function from keras
from keras.models import Sequential
```

```
#importing different layers
from keras.layers import InputLayer,Dense
```

DEFINING THE NUMBER OF INPUT NEURONS

```
xtrain.shape
```

```
(491, 11)
```

```
xtrain.shape[1]
```

```
11
```

```
#defining the input neurons
input_neurons=xtrain.shape[1]
```

```
#defining the no of output neurons
#binary classification problem - output neuron = 1
output_neurons=1
```

```
#defining hidden layer and no of neurons in each layer
number_of_hidden_layers=2
neuron_hidden_layer_1=10
neuron_hidden_layer_2=5
```

```
#defining the architecture of the model
model = Sequential()
model.add(InputLayer(input_shape=(input_neurons,)))
model.add(Dense(units=neuron_hidden_layer_1, activation='relu'))
model.add(Dense(units=neuron_hidden_layer_2, activation='relu'))
model.add(Dense(units=output_neurons, activation='sigmoid'))
```

```
#summary of the model
model.summary()
```

```
Model: "sequential"
```

Layer (type)	Output Shape	Param #
=====	=====	=====
dense (Dense)	(None, 10)	120
dense_1 (Dense)	(None, 5)	55

dense_2 (Dense)	(None, 1)	6
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Total params: 181
 Trainable params: 181
 Non-trainable params: 0

#no of parameter between input and first hidden layer
 $\text{input_neurons} \times \text{neuron_hidden_layer_1}$

110

#Adding bias for each neurons for first hidden layer
 $\text{input_neurons} \times \text{neuron_hidden_layer_1} + 10$

120

#no.of parameter between first and second hidden layer
 $\text{neuron_hidden_layer_1} \times \text{neuron_hidden_layer_2} + 5$

55

#no.of parameter between second hidden layer and output layer
 $\text{neuron_hidden_layer_2} \times \text{output_neurons} + 1$

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COMPILING THE MODEL (DEFINING LOSS AND OPTIMIZER)

```
#compiling the model
# loss as binary_crossentropy, since we have binary classification problem
# defining the optimizer as adam
# Evaluation metric as accuracy
model.compile(loss="binary_crossentropy", optimizer="Adam", metrics=["accuracy"])
```

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