

1 INTRODUCTION

1.1 Overview

Personal loan approval is a process where financial institutions or banks evaluate a borrower's financial status and creditworthiness to decide whether or not to approve their loan application.

Predicting personal loan approval using machine learning involves building a model that can analyze data related to borrowers' demographics, financial history, credit score, and other relevant factors to predict whether or not their loan application will be approved.

Machine learning algorithms, such as logistic regression, decision trees, random forests, and gradient boosting, can be used to analyze historical loan data and identify patterns and relationships that can help predict future loan approvals. By leveraging these algorithms, lenders can make faster and more accurate decisions, resulting in a more efficient and reliable lending process.

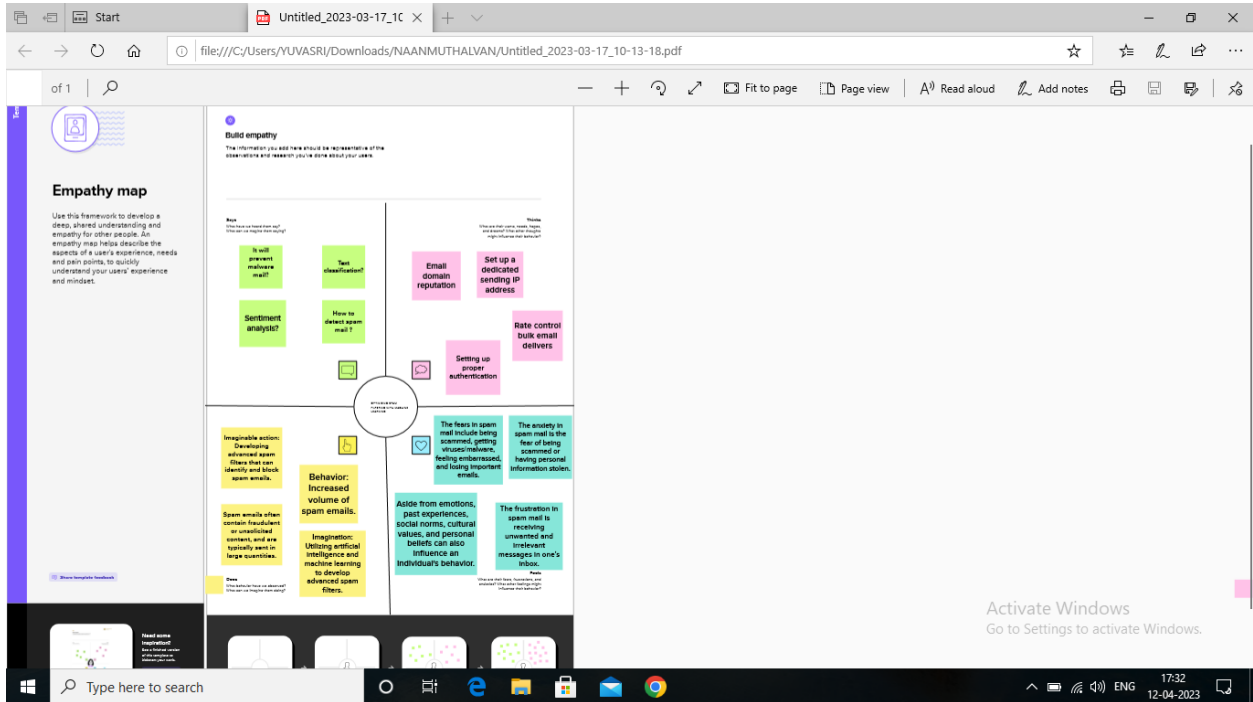
1.2 Purpose

The purpose of predicting personal loan approval using machine learning is to streamline and automate the loan approval process for lenders while improving the accuracy and efficiency of loan decisions.

Traditionally, lenders have used manual underwriting processes that rely on human judgment and are often time-consuming and error-prone

2 problem Definition & design Thinking

2.1 Empathy Map



2.2 Ideation & Brainstorming Map

3. RESULT

The screenshot shows a web application form titled "127.0.0.1:8000/predict". The form contains several input fields and dropdown menus for user data:

- Education:** Graduate (dropdown)
- Self Employed:** No (dropdown)
- Applicant Income:** 5849 (text input)
- CO Applicant Income:** 1234 (text input)
- Loan Amount:** 128 (text input)
- Loan Amount Term:** 360 (text input)
- Credit History:** 1 (text input)
- Property Area:** Urban (dropdown)

At the bottom of the form is a green button labeled "Submit". The form is displayed in a web browser window with a Windows taskbar at the bottom.

127.0.0.1:8000/predict

127.0.0.1:8000/predict

Gmail YouTube Maps

Enter your Details for Loan Approval Prediction

Gender
Male

Married
No

Dependents
0

Education
Graduate

Self Employed
No

Applicant Income
5849

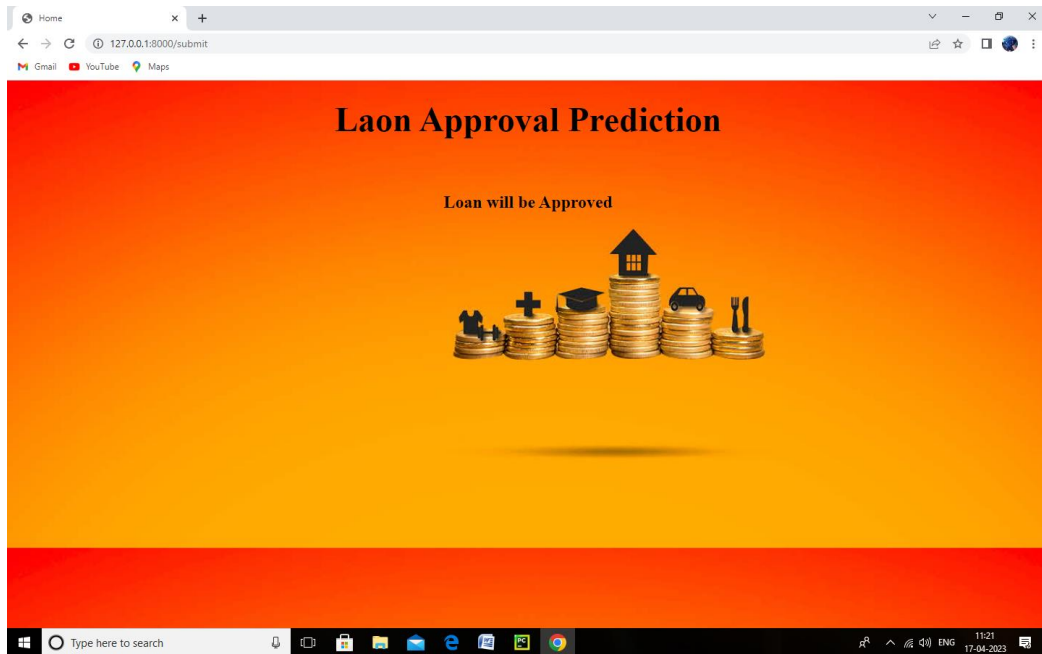
CO Applicant Income
1234

Loan Amount
128

Loan Amount Term

Type here to search

11:21 17-04-2023



3 ADVANTAGES & DISADVANTAGES

Advantages:-

1.Improved accuracy: Machine learning algorithms can analyze vast amounts of data and identify patterns and relationships that may not be apparent to human underwriters. This can lead to more accurate loan decisions and reduce the risk of default.

2.Speed and efficiency: Machine learning algorithms can quickly analyze borrower data and make loan decisions, allowing lenders to process loan applications faster and more efficiently.

Disadvantages:-

1. Biases and discrimination: Machine learning algorithms can perpetuate biases and discrimination if not designed properly. Lenders must ensure that their algorithms are fair and unbiased and do not discriminate against certain groups of borrowers.
2. .Lack of transparency: Machine learning algorithms can be complex and difficult to understand, making it challenging for borrowers to understand why their loan application was denied or approved.

4 APPLICATION

1. Data collection: Lenders need to collect data related to borrowers' demographics, financial history, credit score, and other relevant factors. This data can be obtained from a variety of sources, such as credit bureaus, financial institutions, and government agencies.
2. Data cleaning and preprocessing: Once data is collected, it needs to be cleaned and preprocessed to remove any errors or inconsistencies and make it suitable for analysis.

6 CONCLUSION

1. In conclusion, predicting personal loan approval using machine learning is a powerful tool that can improve the accuracy and efficiency of loan decisions for lenders. Machine learning algorithms can analyze vast amounts of borrower data and identify patterns and relationships that can help predict loan approvals more accurately and quickly than traditional manual underwriting processes.
2. However, there are also challenges associated with predicting personal loan approval using machine learning, such as biases and discrimination, lack of transparency, limited data, and the need for human oversight.

7 FUTURE SCOPE

Improved accuracy and efficiency: As machine learning algorithms continue to improve and evolve, they will become more accurate and efficient in predicting loan approvals, reducing the risk of default and improving the overall efficiency of the lending process.

Integration with blockchain technology: Blockchain technology can provide a secure and transparent way to store and share borrower data, which can enhance the accuracy and efficiency of machine learning algorithms.

8 APPENDIX

A Source code

```
import pandas as pd
```

```
import numpy as np
```

```
import pickle
```

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

```
%matplotlib inline
```

```
import seaborn as sns
```

```
import sklearn
```

```
from sklearn.tree import DecisionTreeClassifier
```

```
from sklearn.ensemble import GradientBoostingClassifier, RandomForestClassifier
```

```
from sklearn.neighbors import KNeighborsClassifier
```

```
from sklearn.model_selection import RandomizedSearchCV
```

```
import imblearn
```

```
from sklearn.model_selection import train_test_split
```

```
from sklearn.preprocessing import StandardScaler
```

```
from sklearn.metrics import accuracy_score, classification_report,  
confusion_matrix, f1_score
```

```
import pandas as pd
```

```
import numpy as np
```

```
data = pd.read_csv('/content/train_u6lujuX_CVtuZ9i.csv')
```

```
data.head()
```

```
data.info()
```

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

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

```
data['Dependents'] = data['Dependents'].str.replace('+', '  
' )
```

```
data['Dependents']=data['Dependents'].fillna(data['Dependents'].mode()[0])
data['Self_Employed']=data['Self_Employed'].fillna(data['Self_Employed'].mode()[0])
data['LoanAmount']=data['LoanAmount'].fillna(data['LoanAmount'].mode()[0])
data['Loan_Amount_Term']=data['Loan_Amount_Term'].fillna(data['Loan_Amount_Term'].mode()[0])
data['Credit_History']=data['Credit_History'].fillna(data['Credit_History'].mode()[0])
```

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

```
data['Gender']=data['Gender'].fillna(data['Gender'].mode()[0])
data['Married']=data['Married'].fillna(data['Married'].mode()[0])
data['Dependents']=data['Dependents'].str.replace('+','')
data['Dependents']=data['Dependents'].fillna(data['Dependents'].mode()[0])
data['Self_Employed']=data['Self_Employed'].fillna(data['Self_Employed'].mode()[0])
```

```
data['LoanAmount']=data['LoanAmount'].fillna(data['LoanAmount'].mode()[0])
data['Loan_Amount_Term']=data['Loan_Amount_Term'].fillna(data['Loan_Amount_Term'].mode()[0])
data['Credit_History']=data['Credit_History'].fillna(data['Credit_History'].mode()[0])
data.info()
data['Gender'].unique()
```



```

data['Gender']=data['Gender'].replace({'Male':1, 'Female':0})
data['Married'].unique()
data['Married']=data['Married'].replace({'Yes':1, 'No':0})

data['Loan_Status']=data['Loan_Status'].replace({'N':0, 'Y':1})

data['Dependents'].unique()
data['Dependents']=data['Dependents'].astype(int)

data['Education'].unique()
data['Education']=data['Education'].replace({'Graduate':1, 'Not Graduate':2})
data['Self_Employed'].unique()
data['Self_Employed']=data['Self_Employed'].replace({'Yes':1, 'No':0})
data['Property_Area'].unique()
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
data['Property_Area']=le.fit_transform(data['Property_Area'])
data['Property_Area']
data.info()
data.describe()
plt.figure(figsize=(12,5))

plt.subplot(121)

sns.distplot(data['ApplicantIncome'], color='r')

plt.subplot(122)

```

```

sns.distplot(data['Credit_History'])

plt.show()
plt.figure(figsize=(18,4))
plt.subplot(1,4,1)
sns.countplot(data['Gender'])
plt.subplot(1,4,2)
sns.countplot(data['Education'])
plt.show()
plt.figure(figsize=(20,5))
plt.subplot(131)
sns.countplot(x=data['Married'], hue=data['Gender'])
plt.subplot(132)
sns.countplot(x=data['Self_Employed'], hue=data['Education'])
plt.subplot(133)
sns.countplot(x=data['Property_Area'], hue=data['Loan_Amount_Term'])
data.head()
import seaborn as sns

sns.swarmplot(x=data['Gender'],y=data['ApplicantIncome'],hue=data['Loan_Status'])
x=data.drop(['Loan_Status','Loan_ID'],axis=1)
y=data['Loan_Status']
x_train, x_test, y_train, y_test = train_test_split(x,
y, test_size=0.33, random_state=42)
x_train
from sklearn.tree import DecisionTreeClassifier
dtc=DecisionTreeClassifier()
dtc.fit(x_train,y_train)
y_pred=dtc.predict(x_test)

```

```
print('***DecisionTreeClassifier***')
print('confusion matrix')
print(confusion_matrix(y_test,y_pred))
print('classification report')
print(classification_report(y_test,y_pred))
import pickle
pickle.dump(dtc,open('rdf.pkl','wb'))
```

app.py

```
import numpy as np

import pickle

import pandas

import os

from flask import Flask, request, render_template

app = Flask(__name__, template_folder='template')

model = pickle.load(open('rdf.pkl', 'rb'))

scale = pickle.load(open('scale1.pkl', 'rb'))
```

```
@app.route('/') # rendering the html template

def home():

    return render_template('home.html') # to access
the home page.
```

```
@app.route('/predict', methods=["POST", "GET"]) #
rendering the html template
```

```
def predict():

    return render_template("input.html")
```

```
@app.route('/submit', methods=["POST", "GET"]) # route
to show the predictions in a web UI
```

```
def submit():

    # reading the inputs given by the user

    input_feature = [int(x) for x in
request.form.values()]

    # input_feature = np.transpose(input_feature)

    input_feature = [np.array(input_feature)]

    print(input_feature)
```

```

    names = ['Gender', 'Married', 'Dependents',
'Education', 'Self_Employed', 'ApplicantIncome',

            'CoapplicantIncome', 'LoanAmount',
'Loan_Amount_Term', 'Credit_History', 'Property_Area']

    data = pandas.DataFrame(input_feature,
columns=names)

    print(data)


# data_scaled = scale.fit_transform(data)

# data = pandas.DataFrame(,columns=names)


# predictions using the loaded model file
prediction = model.predict(data)

print(prediction)

prediction = int(prediction)

print(type(prediction))


if (prediction == 0):

    return render_template("output.html",
result="Loan will not be Approved")

else:

```

```
        return render_template("output.html",
result="Loan will be Approved")

    # showing the prediction results in a UI


if __name__ == "__main__":

    app.run(host='0.0.0.0', port=8000, debug=True) #
running the app

    port = int(os.environ.get('PORT', 5000))

    app.run(debug=False)
```

home.html

```
<!doctype html>

<html lang="en">

    <head>

        <!-- Required meta tags -->

        <meta charset="utf-8">

        <meta name="viewport" content="width=device-width,
initial-scale=1">
```

```

<!-- Bootstrap CSS -->

<link
href="https://cdn.jsdelivr.net/npm/bootstrap@5.0.0-
beta3/dist/css/bootstrap.min.css" rel="stylesheet"
integrity="sha384-
eOJMYsd53ii+scO/bJGFsiCZc+5NDVN2yr8+0RDqr0Ql0h+rP48ckx1
pbzKgwra6" crossorigin="anonymous">

<link
href="https://unpkg.com/tailwindcss@^2/dist/tailwind.mi
n.css" rel="stylesheet">

<title>loan Prediction</title>

</head>

<body>

<!-- This example requires Tailwind CSS v2.0+ -->

<div class="relative bg-white overflow-hidden">

  <div class="max-w-7xl mx-auto">

    <div class="relative z-10 pb-8 bg-white sm:pb-16
md:pb-20 lg:max-w-2xl lg:w-full lg:pb-28 xl:pb-32">

      <svg class="hidden lg:block absolute right-0
inset-y-0 h-full w-48 text-white transform translate-x-
1/2" fill="currentColor" viewBox="0 0 100 100"
preserveAspectRatio="none" aria-hidden="true">

        <polygon points="50,0 100,0 50,100 0,100" />

```

```
</svg>
```

```
<div class="relative pt-6 px-4 sm:px-6 lg:px-8">
```

```
  <nav class="relative flex items-center justify-between sm:h-10 lg:justify-start" aria-label="Global">
```

```
    <div class="flex items-center flex-grow flex-shrink-0 lg:flex-grow-0">
```

```
      <div class="flex items-center justify-between w-full md:w-auto">
```

```
        <a href="#">
```

```
          <span class="sr-only">Workflow</span>
```

```
          <img class="h-8 w-auto sm:h-10" src="">
```

```
        </a>
```

```
      <div class="-mr-2 flex items-center md:hidden">
```

```
        <button type="button" class="bg-white rounded-md p-2 inline-flex items-center justify-center text-gray-400 hover:text-gray-500 hover:bg-gray-100 focus:outline-none focus:ring-2 focus:ring-inset focus:ring-indigo-500" aria-expanded="false">
```

```
          <span class="sr-only">Open main menu</span>
```

```
        <!-- Heroicon name: outline/menu -->
```



```
        <svg class="h-6 w-6"
xmlns="http://www.w3.org/2000/svg" fill="none"
viewBox="0 0 24 24" stroke="currentColor" aria-
hidden="true">

        <path stroke-linecap="round"
stroke-linejoin="round" stroke-width="2" d="M4 6h16M4
12h16M4 18h16" />

        </svg>

    </button>

</div>

</div>

</div>

<div class="hidden md:block md:ml-10 md:pr-4
md:space-x-8">

    <a href="#" class="font-medium text-gray-
500 hover:text-gray-900">Home</a>

    <a href="#" class="font-medium text-gray-
500 hover:text-gray-900">Prediction</a>

    <a href="#" class="font-medium text-gray-
500 hover:text-gray-900">About us</a>
```

```
        <a href="#" class="font-medium text-gray-500 hover:text-gray-900">contact</a>
```

```
    </div>
```

```
</nav>
```

```
</div>
```

```
<!--
```

```
    Mobile menu, show/hide based on menu open state.
```

```
    Entering: "duration-150 ease-out"
```

```
        From: "opacity-0 scale-95"
```

```
        To: "opacity-100 scale-100"
```

```
    Leaving: "duration-100 ease-in"
```

```
        From: "opacity-100 scale-100"
```

```
        To: "opacity-0 scale-95"
```

```
-->
```

```
    <div class="absolute top-0 inset-x-0 p-2 transition transform origin-top-right md:hidden">
```

```

        <div class="rounded-lg shadow-md bg-white
ring-1 ring-black ring-opacity-5 overflow-hidden">

            <div class="px-5 pt-4 flex items-center
justify-between">

                <div>

                </div>

                <div class="-mr-2">

                    <button type="button" class="bg-white
rounded-md p-2 inline-flex items-center justify-center
text-gray-400 hover:text-gray-500 hover:bg-gray-100
focus:outline-none focus:ring-2 focus:ring-inset
focus:ring-indigo-500">

                        <span class="sr-only">Close main
menu</span>

                        <!-- Heroicon name: outline/x -->

                        <svg class="h-6 w-6"
xmlns="http://www.w3.org/2000/svg" fill="none"
viewBox="0 0 24 24" stroke="currentColor" aria-
hidden="true">

                            <path stroke-linecap="round"
stroke-linejoin="round" stroke-width="2" d="M6 18L18
6M6 6l12 12" />

```

</svg>

</button>

</div>

</div>

<div class="px-2 pt-2 pb-3 space-y-1">

Home

prediction

about us

contact

</div>

</div>

</div>

<main class="mt-10 mx-auto max-w-7xl px-4 sm:mt-12 sm:px-6 md:mt-16 lg:mt-20 lg:px-8 xl:mt-28">

<div class="sm:text-center lg:text-left">

<h1 class="text-4xl tracking-tight font-extrabold text-gray-900 sm:text-5xl md:text-6xl">

Loan Prediction

Machine Learning

</h1>

<p class="mt-3 text-base text-gray-500 sm:mt-5 sm:text-lg sm:max-w-xl sm:mx-auto md:mt-5 md:text-xl lg:mx-0">

Lorem ipsum dolor sit amet consectetur adipisicing elit. Excepturi ad perspiciatis dolores, deleniti culpa odit dolorem harum dolore ex amet.

</p>

<div class="mt-5 sm:mt-8 sm:flex sm:justify-center lg:justify-start">

<div class="rounded-md shadow">

<a href="/predict" class="w-full flex items-center justify-center px-8 py-3 border border-

transparent text-base font-medium rounded-md text-white
bg-indigo-600 hover:bg-indigo-700 md:py-4 md:text-lg
md:px-10">

Prediction

</div>

</div>

</div>

</main>

</div>

</div>

<div class="lg:absolute lg:inset-y-0 lg:right-0 lg:w-
1/2">

</div>

</div>

```
    <!-- Option 1: Bootstrap Bundle with Popper -->

    <script
src="https://cdn.jsdelivr.net/npm/bootstrap@5.0.0-
beta3/dist/js/bootstrap.bundle.min.js"
integrity="sha384-
JEW9xMcG8R+ph31jmWH6WWP0WintQrMb4s7ZOdauHnUtxwoG2vI5DkL
tS3qm9Ekf" crossorigin="anonymous"></script>

    </body>

</html>
```

Input.html

```
<!DOCTYPE html>

<html xmlns="http://www.w3.org/1999/html">

<head>

    <style>

        /* Style inputs with type="text", select elements
and textareas */
```

```
input[type=number], select, textarea {  
    width: 100%; /* Full width */  
    padding: 12px; /* Some padding */  
    border: 1px solid #ccc; /* Gray border */  
    border-radius: 4px; /* Rounded borders */  
    box-sizing: border-box; /* Make sure that  
padding and width stays in place */  
    margin-top: 6px; /* Add a top margin */  
    margin-bottom: 16px; /* Bottom margin */  
    resize: vertical /* Allow the user to vertically  
resize the textarea (not horizontally) */  
}
```

```
/* Style the submit button with a specific  
background color etc */
```

```
input[type=submit] {  
    background-color: #04AA6D;  
    color: white;  
    padding: 12px 20px;  
    border: none;  
    border-radius: 4px;
```



```
        cursor: pointer;

    }

    /* When moving the mouse over the submit button,
    add a darker green color */

    input[type=submit]:hover {

        background-color: #45a049;

    }

    /* Add a background color and some padding around
    the form */

    .container {

        border-radius: 5px;

        background-color: #f2f2f2;

        padding: 20px;

    }

body {

    background-size: cover;
```

```
background-image: url('https://encrypted-  
tbn0.gstatic.com/images?q=tbn:ANd9GcR70TDTAUcRk7Q7W2nK-  
aIqsmoTN6VruMi0mA&usqp=CAU');
```

```
}
```

```
</style>
```

```
</head>
```

```
<body>
```

```
<h3>Enter your Details for Loan Approval  
Prediction</h3>
```

```
<div class="container">
```

```
<form action = '/submit', method = 'post'>
```

```
<label for="Gender">Gender</label>
```

```
<select id="Gender" name="Gender">
```

```
<option value=0>Male</option>
```

```
<option value=1>Female</option>
```

```
</select>
```

```
<label for="Married">Married</label>
```

```
<select id="Married" name="Married">
```

```
  <option value=1>Yes</option>
```

```
  <option value=0>No</option>
```

```
</select>
```

```
<label for="Dependents">Dependents</label>
```

```
  <input type="number" id="Dependents" min = 0 max =  
10 name="Dependents" placeholder="No of Dependents on  
you.....">
```

```
<label for="Education">Education</label>
```

```
<select id="Education" name="Education">
```

```
  <option value=1>Graduate</option>
```

```
  <option value=0>Not Graduate </option>
```

```
</select>
```

```
<label for="Self_Employed">Self Employed</label>

<select id="Self_Employed" name="Self_Employed">

  <option value=1>Yes</option>

  <option value=0>No</option>

</select>
```

```
<label for="ApplicantIncome">Applicant
Income</label>
```

```
<input type="Number" min = 1000
id="ApplicantIncome" name="Applicant Income"
placeholder="Your Income...">
```

```
<label for="CoapplicantIncome">CO Applicant
Income</label>
```

```
<input type="Number" min = 100
id="CoapplicantIncome" name="Co Applicant Income"
placeholder="Your Co Applicant Income...">
```

```
<label for="LoanAmount">Loan Amount</label>
```

```
<input type="Number" min = 0 id="LoanAmount"
name="Loan Amount" placeholder="Enter the Loan Amount
...">
```

```
<label for="Loan_Amount_Term">Loan Amount  
Term</label>
```

```
<input type="Number" min = 30 max = 15000  
id="Loan_Amount_Term" name="Loan Amount Term"  
placeholder="Enter the Term Loan Amount in days ...">
```

```
<label for="Credit_History">Credit History</label>
```

```
<input type="Number" min = 0 max = 5  
id="Credit_History" name="Credit History"  
placeholder="Enter the Your Previous Credit History  
...">
```

```
<label for="Property_Area">Property Area</label>
```

```
<select id="Property_Area" name="Property Area">
```

```
<option value=2>Urban</option>
```

```
<option value=0>Rural</option>
```

```
<option value=1>Semi Urban</option>
```

```
</select>
```

```
<input type = 'submit' value = 'Submit'>
```

```
</div>
```

```
</body>
```

```
</html>
```

Output.html

```
<!DOCTYPE html>
```

```
<html>
```

```
<head>
```

```
<title>Home</title>
```

```
<style>
```

```
body
```

```
{
```

```
    background-image: url("https://www.dbs.com/in/iwov-  
resources/media/images/learn/banners/what-is-a-  
personal-loan-1404x630.jpg");
```

```
        background-size: cover;

    }

    .pd{

padding-bottom:45%;}

    }

</style>

</head>

<body>

<form action="/submit" method="post">

<br>

<center><b class="pd"><font color="black" size="15"
font-family="Comic Sans MS" >Laon Approval
Prediction</font></b></center><br><br>

<div>

<br>

<center>

<h2><font color="black"> {{result}} </h2>

</center>

</div>

</form>
```

</body>

</html>