# SMART LENDER -APPLICANT CREDIBILITY PREDICTION FOR LOAN APPROVAL

**TEAM ID :PNT2022TMID29739** 

# A PROJECT REPORT

Submitted by

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> in partial fulfillment for the award of the degree Of

# BACHELOR OF ENGINEERING I N ELECTRONICS AND COMMUNICATION ENGINEERING

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GitHub and Project Demo Link

# 1 .INTRODUCTION

# 1.1 PROJECT OVERVIEW

The Smart lender project is used to provide a way for the bank to trust the customer and provide the loan to a trustworthy person. This project decreases the approval time and decreases the risk associated with the loan. The aim of this project was to compare the various loan prediction models and show which is the best one with least amount of error and could be used by banks in real world to predict if the loan should be approved or not taking the risk factor in mind. After analyzing all models Random forest model is selected as the best model for predicting the loan eligibility of the customer

#### 1.2 PURPOSE

The smart lender project is used to predict whether a particular applicant is safe to provide loan

- The entire process of verifying the customer characteristics will be automated by machine learning technology.
- This credit forecasting is very useful for both bank employees and customer
- The customer can also provide their feedback in this system which helps the bank to improve their service

# 2 LITERATURE SURVEY

# **2.1** EXISTING PROBLEM

Bank employees check the details of applicant and give the loan to eligible a pplicant. Checking the details of all applicant takes a lot of time. Assessing the risk, which is involved in a loan application, is one of the most important concerns of the banks for survival in the highly competitive market and for profit ability. These banks receive number of loan applications for their customers and other people on daily basis. Not everyone gets approved. Most of the banks use their own credit scoring and risk assessment techniques in order to analyze the loan application and to make decisions on credit approval. In spite of this, there are many cases happening every year, where people do not repay the loan amount or they default, due to which these financial institution suffer huge amount of losses.

#### 2.2 REFERENCE

# Paper 1: An Approach For Prediction Of Loan Approval

Publication year: May-June 2021

Author Name: Ms. Kathe Rutika Pramod

**Summary:** In our banking system banks have many products to sell but main source of income of any banks is on its credit line they can earn from interest of those loans which they credits a bank's profit or a loss depends to a large extent on loans the customers are paying back the loan defaulting by predicting the loan defaulters.

**Methodology used:** The prediction model which is constructed using three different training algorithms to train a supervised two layer feedforward network. The results show that the training algorithm improves the design of loan default prediction model.and we use of machine learning

# Paper 2: Loan Prediction System using Machine Learning

**Publication year:**March-April 2021 **Author name:**Ramrao ,yash patil

Journal name: International Conference on Advances in Computing and

Communication (ICACC)

**Summary**: As the needs of people are increasing, the demand for loans in banks is also frequently getting higher every day. Banks typically process an applicant's loan after screening and verifying the applicant's eligibility, which is a difficult and time-consuming process. In some cases, some applicants default and banks lose capital. The Machine Learning Approach is ideal for reducing human effort and effective decision making in the loan approval process by implementing machines.

**Methodology used**: The algorithm which will be used for data modeling is Logistic Regression using stratified k-folds cross-validation and Random Forest.

# Paper 3:Loan Credibility Prediction System Based on Decision Tree Algorithm

**Publication year**:09 September 2015

**Author name:** Sivasree M S, Rekha Sunny T

**Journal Name:** International Journal of Engineering Research & Technology (IJERT)

**Summary:** Data mining techniques are becoming very popular now-a-days because of the wide availability of huge quantities of data and the need for transforming such data into knowledge. Techniques of data mining are implemented in various domains such as retail industry, telecommunication industry, biological data analysis, intrusion detection and other scientific applications. Data mining techniques can also be used in the banking industry which help them compete in the market well equipped. In this paper we introduce an effective prediction model for the bankers that helps them predict the credible customers who have applied for loan. Data Mining Algorithm is applied to predict the attributes relevant for credibility.

**MethodologyUsed**: DecisionTree, CreditRiskAssessment, Classification, Prediction, Attribute Selection.

# <u>Paper 4:Rethinking SME default prediction: a systematic literature review and future perspectives</u>

Publication year: 29 January 2021

Author name: Ciampi, F., Giannozzi, A., Marzi, G.

Journal Name: Scientometrics

**Summary:** Over the last dozen years, the topic of small and medium enterprise (SME) default prediction has developed into a relevant research domain that has grown for important reasons exponentially across multiple disciplines, including finance, management, accounting, and statistics.

# Paper 5: Analysis of Loan Availability using Machine Learning Techniques

**Publication year:** September 2021

Author name: Sharayu Dosalwar, Ketki Kinkar, Rahul Sannat, Dr Nitin Pise

JournalName: International Journal of Advanced Research in Science,

Communication and Technology (IJARCET)

**Summary:** In the banking system, banks have a variety of products to provide, but credit lines are their primary source of revenue. As a result, they will profit from the interest earned on the loans they make.Loans, or whether customers repay or default on their loans, affect a bank's profit or loss. The bank's Non-Performing Assets will be reduced by forecasting loan defaulters. As a result, further investigation into this occurrence is essential. Because precise forecasts are essential for benefit maximization, it's crucial to analyze and compare the various methodologies

# <u>Paper 6:Analysis and Comparison of Loan Sanction Prediction Model Using Python</u> <u>Publication year: Jun 2018</u>

**Author name:** SRISHTI SRIVASTAVA, AYUSH GARG, ARPIT SEHGAL & ASHOK KUMAR **JournalName:**International Journal of Computer Science Engineering and Information Technology Research (IJCSEITR)

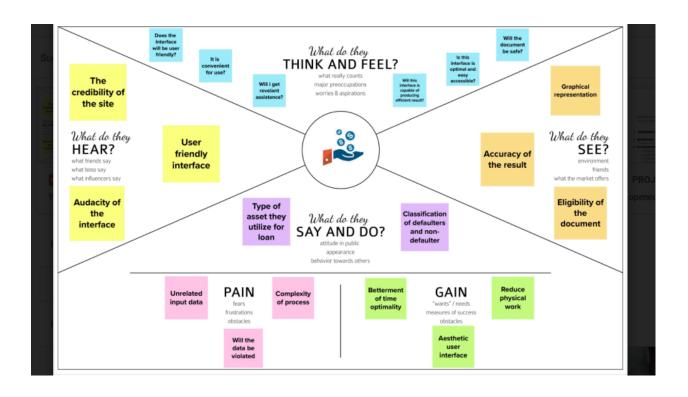
**Summary:** Sanctioning of loan to borrowers form the most vital part of every bank's business, as most of its assets come from the profit gained in the loan distribution process. Therefore, it is essential for banks to estimate whether the customer is right.

# 2.3 PROBLEM STATEMENT DEFINITION:

A bank is a financial institution licensed to receive deposits and make loans needs a way to verify the customer details and their documents for getting loan because they need a trustworthy customer with proper credentials who can repay the loan amount and interest on tim

#### 3 .IDEATION & PROPOSED SOLUTION

#### 3.1 EMPATHY MAP CANVAS

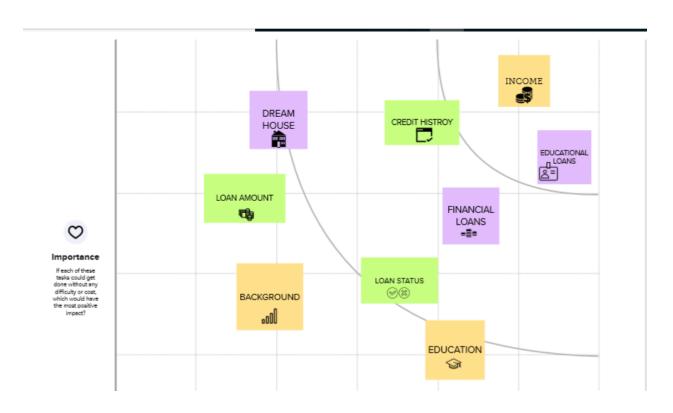


#### 3.2 IDEATION & BRAINSTORMING:

Define your problem statement:

A bank is a financial institution licensed to receive deposits and make loans needs a way to verify the customer details and their documents for getting loan because they need a trustworthy customer with proper credentials who can repay the loan amount and interest on time

Person 1			Person	2		Person 3		
Automate loan process	loan status have two values	YES OR NO	NO of dependent	Today target amount	Business standing	ELIGIBLITY	BINARY	MULTI CLASS
approved or not approved	online application form filling	INCOME	DEPLOY MENT	Evaluation	MODELING	DEAM HOUSE FINANCIAL	SEMI URBAN	RURAL
LOAN AMOUNT	CREDIT HISTORY	EDUCATION	DATA UNDER STANDING	LOT OF MAN HOURS SAVE	CUSTOMER SEGMENTS	ALL URBAN	REAL TIME	CUSTOMER DETAILS



Prioritizing ideas

**10** 

# 3.3 PROPOSED SOLUTION

SL.NO	PARAMETER	DESCRIPTION
1	Problem Statement (Problem to be solved)	* They have a presence across all urban ,semi urban, eligibility of loan, income,education
		*Applicants with high income should have more chances of getting approval.less the time period has higher chances of approval
2	Idea / Solution description	To gather loan data from multiple data sources and use various machine learning algorithms  * Who has paid their historical debts have more chances of getting approval.less the amount higher changes
3	Novelty / Uniqueness	*Our contribute the work lies in deep learning.we want to use (BIG DATA ALGORITHM)  *They provide complete end-to-end machine learning workflow for building we includes methods like automated process in future

#### 3.4 PROBLEM SOLUTION FIT

J&P

1. CUSTOMER SEGMENT(S)
Our target customers are mostly banking firm, small financial firm that lends out loan and credit card companies because of the increasing rate of loan defaulter and also to increase the slow process of the loan approval.

#### 6. CUSTOMER CONSTRAINTS

Banks are not to correctly handle the loan request. People within a protected class being clearly treated differently than those of non-protected classes for loan. There is an increasing rate of loan defaults. Banks Identify the loan defaulters for much-reduced credit risk as large portions of a bank's assets directly come from the interest earned on loans given.

#### 5. AVAILABLE SOLUTIONS

- Random forest, Logistic regression, Decision tree and Naive bayes algorithm are used
- Using data pre-processing data mining and data filtering
- Algorithms such as naïve bayes, k-nearest neighbors

2. JOBS-TO-BE-DONE / PROBLEMS

Needs to Support genuine Entrepreneur That the process should be easier a timesaving. To find an applicant which can give best interest. Needs to find a loan applicant with good credit score

#### 9. PROBLEM ROOT CAUSE

The root cause of this problem is the banks identify the loan defaulters formuch-reduced credit risk as large portions of a bank's assets directly come from the interest earned on loans given. People within a protected class being clearly treated differently than those of non-protected classes for loan

#### 7. BEHAVIOUR

RC

Directly related:

The customers who lends the loan and the banks that checks the credibility seek to do the process faster.

#### Indirectly associated:

The small finance sector that deals with middle class and poor class people seek to find the credibility.

BE

AS

# 4 .REQUIREMENT ANALYSIS

# **4.1 FUNCTIONAL REQUIREMENT**

USER. No	Functional Requirement	subRequirement (Story/Sub T ask)
usr-1	User Registration	Registration through Form e-mail
usr-2	User Login	Login using username and password
usr-3	Bank Login	LOGIN using bank credentials
usr-4	User View Profile	User will find the complete details and steps for applying loan
usr-5	User Application	User upload their details and document proof
usr-6	Contact	User can contact the bank in case of any queries

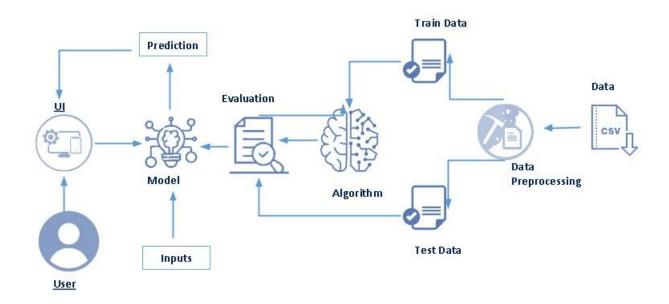
# 4.2 NON-FUNCTIONAL REQUIREMENT

following are the non-functional requirements of the proposed solution

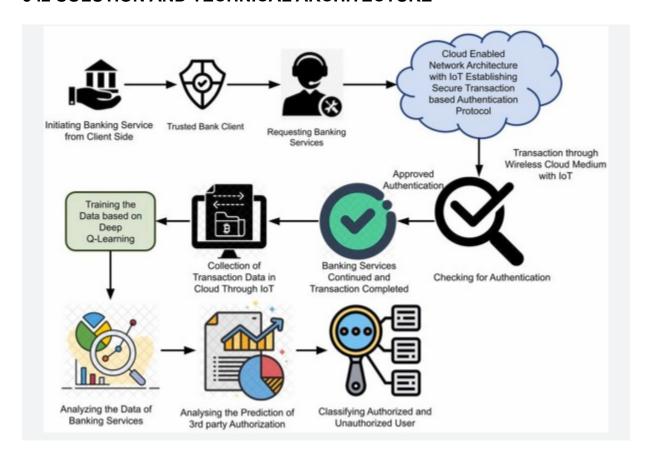
USER NO	NON-FUNCTIONAL REQUIREMENT	DESCRIPTION
User-1	Usability	User friendly U1 Easy to use
user-2	Security	Login Verification User details not shared to third party Verification by bank Captcha security
User -3	Reliability	Handle changes
user-4	Performance	Responsive Website Quick loading
user-5	Availability	Good Services Availability Satisfies user needs Meets user requirements
user-6	Scalability	The System Can Grow Without negative influence Adaptable Handle multiple users64

# **5.PROJECT DESIGN**

# **5.1 DATA FLOW DIAGRAM**



#### **5.2 SOLUTION AND TECHNICAL ARCHITECTURE**



# **Application characteristics**

S.no	Characteristics	Description	Technology
1	Open-Source FrameworksList	Listhe open-source frameworks used	IBM Cloud
2	Security Implementations	ItsSecures information and data	IBM Cloud provides layered security controls across network and infrastructure
3	Scalable Architecture	Its support various data sizes	IBM Cloud
4	Availability	Creating multiple pages for comfortable user interface experience	HTML,CSS, JavaScrip

# **5.3 USER STORIES**

User Type	Functional Requirement	User Story no	User story/ Task	Priority	Priority Release
customer	Registration	USN-1	As a user,I can register for the application by entering my name,email Passwords etc.	HIGH	Sprint-1
	Login	USN-2	As a user,I can log into the application by entering the user's name and password	HIGH	Sprint-1
	Navigation	USN-3	As a user,I can Navigate to different tabs like home,descript -ion,contact	LOW	SPrint-2
	View procedure	USN-4	As a user,I can View the procedure to apply for loan	MEDIUM	Sprint-3
	Contact	USN-5	As a user,I can Contact bank	HIGH	Sprint-3

# 6.PROJECT PLANNING AND SCHEDULING

# **6.1 SPRINT PLANNING AND ESTIMATION**

Sprint	Functional Requiremen t (Epic)	User Story Number	User Story / Task Sprint	Story Points	Priority	Team Members
Sprint-1	Login	USN-1	As a user, I can log into the application by entering the username and password	2	HIGH	JANANI K GOKULNATH
Sprint-1		USN-2	As a user login the gmail id	3	HIGH	NALIN KUMAR JANANI
Sprint-2		USN-3	As a user, I can view the procedure to apply for loan	2	LOW	GOKULNATH NALINKUMAR
Sprint-2		USN-4	AS a user upload the document files for loan verification	2	MEDIUM	ANADHASAYAN GOKULNATH
Sprint-3		USN-5	As a Bank administrator, I can Approve/Reject the loan for the customer based on their details.	1	HIGH	ANADHASAYAN NALINKUMAR
Sprint-4		USN-6	As a user, I can get confirmation of loan approval through email	2	HIGH	GOKULNATH JANANI

#### 6.2 SPRINT DELIVERY SCHEDULE

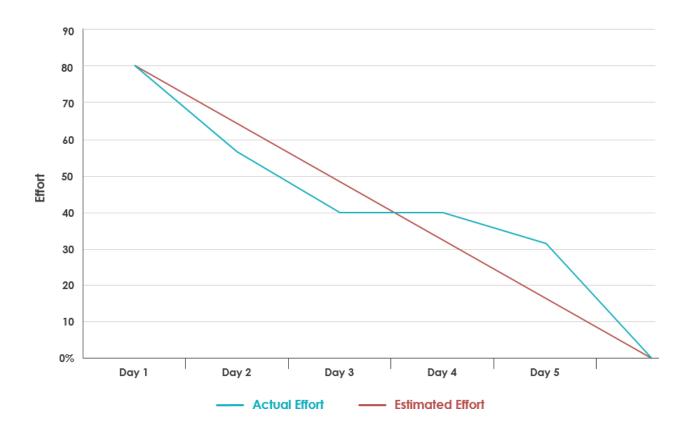
Sprint	Total Story Points	Duration	Sprint End Date	Sprint End Date (Planned)	Story Points Complete	Sprint Release Date
Sprint-1	5	6 days	24 oct 2022	29 oct 2022	5	29 oct 2022
Sprint-2	6	6 days	31 oct 2022	05 nov 2022	7	05 nov 2022
Sprint-3	7	6 days	07 nov 2022	12 nov 2022	7	09 nov 2022
Sprint-3	8	6 days	14 nov 2022	19 nov 2022	6	17 nov 2022

$$\begin{array}{ccc} AV = & \underline{Sprint\ duration} & = & \underline{20} = & 2 \\ Velocity & & 10 \end{array}$$

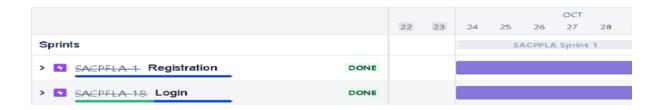
AVERAGE VELOCITY =6/6=1

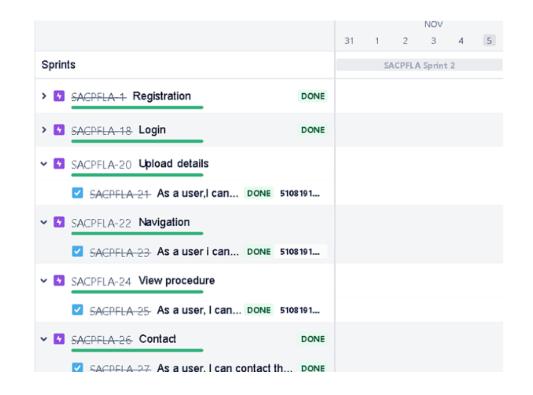
# **Burndown Chart**

A burndown chart shows the amount of work that has been completed in an epic or sprint, and the total work remaining.



# **6.3 REPORTS FROM JIRA**





# 7. CODING AND SOLUTIONING

# 7.1 Feature

# 1:Contact

The user can contact the bank in case of any queries through this contact feature . They should fill in their name, email ID , mobile number and the message they want to send to them.

#### CODE:

home.html

```
C: > Users > ELCOT > Desktop > ibmmmm > templates > ♦ home.html > ♦ style > 😘 nav .teams
       .teams ul li{
            margin-left: 8px;
            color: \Box rgb(\theta, \theta, \theta);
            font-family:system-ui, -apple-system, BlinkMacSystemFont, 'Segoe UI', Roboto, Oxygen, Ubuntu, Cantarell, 'Open Sans', 'Helvetica Neue', sans-
            font-weight: 500;
background: ■white;
            padding: 8px 15px;
            border-radius: 30px;
transition: all 0.3s ease;
        .teams ul li a:hover{
  background: □rgb(0, 16, 49);
  color: ■white;
            background: url('https://static.vecteezy.com/system/resources/previews/005/085/280/original/a-new-startup-has-successfully-applied-for-a-loar
            height: 95vh;
            width: 100%;
            background-size: cover;
            background-position: center;
            position: relative:
            height: 100%;
width: 100%;
```

#### Home css.

```
left: 50%;
   width: 100%;
   padding: 0 20px;
   text-align: center;
   color: □rgb(0, 20, 31);
   font-size: 55px;
   font-weight: 900;
   font-family: 'Gill Sans', 'Gill Sans MT', Calibri, 'Trebuchet MS', sans-serif;
.center .sub_title{
color: □rgb(5, 21, 65);
   font-weight: 900;
   font-family:'Gill Sans', 'Gill Sans MT', Calibri, 'Trebuchet MS', sans-serif;
   margin-top: 20px;
  height: 45px;
   border-radius: 25px;
   border: none;
   margin: 0 10px;
   border: 1px solid ■white;
```

# 7.2 FEATURE

# 2:Prediction

The user will upload all their details in this prediction page to predict the eligibility of them to to get the loan

# **CODE:**

# **Prediction.html**

```
.wrapper .form{
 width: 100%;
.wrapper .form .inputfield{
  margin-bottom: 15px;
 align-items: center;
.wrapper .form .inputfield label{
 width: 200px;
  color: □#757575;
   margin-right: 10px;
  font-size: 14px;
 width: 100%;
  border: 1px solid ■#d5dbd9;
  padding: 8px 10px;
  border-radius: 3px;
transition: all 0.3s ease;
.wrapper .form .inputfield .textarea{
  width: 100%;
  height: 125px;
  resize: none:
```

```
width: 100%;
 height: 125px;
.wrapper .form .inputfield .custom_select{
 position: relative;
 width: 100%;
 height: 37px;
.wrapper .form .inputfield .custom_select:before{
  content: "";
  position: absolute;
 top: 12px;
 right: 10px;
 border: 8px solid;
 border-color: ■#d5dbd9 transparent transparent transparent;
 -webkit-appearance: none;
 -moz-appearance: none;
  appearance:
  width: 100%;
  height: 100%;
  border: 0px;
  padding: 8px 10px;
```

```
| width: 100%; | padding: 8px 10px; | fort-size: 15px; | border: 0px; | border-padics: 3px; | outline: none; | border-padics: 3px; | outline: none; | list | list | outline: none; | list |
```

# **Prediction css:**

```
| Clabel>Co Applicant Income</label>
| Cinjut type="number" class="input" name="CoapplicantIncome" min="0">
| Cidiv Class="inputfield">
| Cidiv Class="custom_select">
|
```

#### 7.3 FEATURE

# 3: Rating and Reviews

The user can provide their feedback at last about the service so that the bank can improve their service

# **CODE:**

```
<!DOCTYPE html>
<html lang="en" dir="ltr">
    <meta charset="utf-8">
     <title>Loan approva status</title>

<
    <h1>LOAN APPROVAL STATUS</h1>
     <h2>{{prediction_text}}</h2>
<img src="static/approve.jpg" width="150px" height="150px">
<h3>Please provide your feedback</h3>
     <div class="container">
        <div class="post">
          <div class="text">Thanks for rating us!</div>
           <div class="edit">EDIT</div>
        <div class="star-widget">
          <input type="radio" name="rate" id="rate-5">
          <label for="rate-5" class="fas fa-star"></label>
<input type="radio" name="rate" id="rate-4">
           <input type="radio" name="rate" id="rate-3">
           <input type="radio" name="rate" id="rate-2";</pre>
           <label for="rate-2" class="fas fa-star"></label>
           <input type="radio" name="rate" id="rate-1">
           <label for="rate-1" class="fas fa-star"></label>
```

```
C: > Users > ELCOT > Desktop > Ibm project > templates > (*) approve.html > (*) html > (*) body > (*) h3
                 <div class="textarea">
                   <textarea cols="30" placeholder="Describe your experience.."></textarea>
                 <div class="btn">
                  <button type="submit">Post</button>
           <script>
             const btn = document.querySelector("button");
             const post = document.querySelector(".post");
             const widget = document.querySelector(".star-widget");
             const editBtn = document.querySelector(".edit");
             btn.onclick = ()=>{
               widget.style.display = "none";
               post.style.display = "block";
               editBtn.onclick = ()=>{
                 widget.style.display = "block";
                 post.style.display = "none";
           </script>
```

# 8.TESTING

# 8.1 TEST CASES

Test case ID	Functional	Test Scenario	Test data	Result	Status
Home page_TC_oo1	Functional	Verify user is able to see the Home page when the user enters the url	url of the web application	Home page should be displayed	
Home page_TC_oo2	UI	Verify the UI elements in the home page	home.html	Application should display UI elements: a.home tab b.about tab c.procedure tab d.user login tab e.predict button	pass
Home page_TC_oo3	functional	Verify user is able to see the Login page when user clicked on User 1 ogin butt	home.html	Login page should be display	pass
Home page_TC_oo4	functional	Verify user is able to log into application with valid credentials	Username: mm@gmail.co m password: mohit	User should navigate to prediction page	pass

Home page_TC_oo5	functional	Verify user is able to log into application with invalid credentials	Username:mm /EmailId: a ss@gmail.com password: mohit	Application should show ' Login failed Invalid username or email ID or password' validation message	pass
prediction page_TC_oo6	functional	Verify user is able to see the prediction page when user clicked on predict button	prediction.html	Prediction page should be display	pass
prediction page_TC_oo7	UI	Verify the UI elements in prediction page	prediction.html	Application should show below U I elements: a . Name text box b . Email ID text box c .Gender drop down menu d .Education drop down menu e.Self_Employ ed drop down menu f.married drop down menu g.Dependents h.property area drop down menu i.credit History jApplicant income	Pass

				kCo Applicant income ILoan amount text . Predict button	
prediction page_TC_oo9	functional	Verify user is able to predict the results with Valid credentials	Name:aishu E mail ID:aishu@gma il.com Gender:Femal e Education: Graduate Self-Employed :No Married:No Dependents:3+ Property Area: Urban CreditHistory: No Applicant Income:10000 Co-applicant Incom-e:5000 Loan amount:7000 Loan amount term:3600 Property document:proo f	Application should be directed to the approve page or reject p age based on their given data	Pass
Approve page_TC_010	functional	Verify user is able to see the Approve page when user clicked on predict button and when predictive value	approve.html	Approve page should be display	pass

## 8.2 USER ACCEPTANCE TESTING

# **Defect Analysis**

Resolution	Severity 1	severity2	severity3	severity4	subtotal
By Design	7	2	1	1	11
Duplicate	1	0	0	0	1
External	2	3	0	1	6
Fixed	1	2	4	12	19
Not Reproduced	0	0	1	0	1
Skipped	0	0	1	1	2
Won"t Fix	0	0	2	1	3
Totals	11	7	9	16	43

# **Test Case Analysis**

section	<b>Total Cases</b>	Not Tested	Fail	Pass
Login	5	0	0	5
Prediction	3	0	0	3
Rating	1	0	0	1

# 9 .RESULTS

## 9 .1 PERFORMANCE METRICS

S.no	Para meter	Values	Screenshot
1	meteric es	Classification model Confusion matrix	. *** Random Forest Classifier ***  Confusion Matrix [[SI 9]
		0-[[51 9]] 1-[[7 72]] Accuracy score Random Forest: Testing accuracy = 0.777292576419214 Training Accuracy= 0.9891774891774892	1 0,89 0.91 0.90 79 accuracy 0.88 139 micro ang 0.88 0.88 139 welchted and 0.88 0.88 0.88 139 welchted and 0.88 0.88 0.88 139  **Trick **min **, trick **, t
		Precision recall fi-score support 0 0.77 0.75 0.76 106	hooding errory is antiferrorised
		1 0.79 0.80 0.80 123	
		Accuracy 0.78 229 Macro avg 0.78 0.78 0.78 229	
		Weighted 0.78 0.78 0.78 229	

#### 10.ADVANTAGES AND DISADVANTAGES

#### **ADVANTAGES**

The customer can predict their eligibility from any part of the world and at any time so it provides user convenience

- Eligible applicant will be sanctioned loan without any delay
- Minimal documentation is required and there is no physical submission of documents
- Whole process will be automated, so human error will be avoided
- Time period for loan sanctioning will be reduced and more Accurate prediction for loan eligibility will be given
- . The customer can contact bank at any time in case of any queries and we had also provided the detailed procedure for applying loan and customer can also provide the rating

#### DISADVANTAGES

- The customer can contact the lender only through online using email or call them in case of any queries
- The bank should externally connect to database and use this software in real time we had provided only the feature
- There may be some risk associated with security of the customers as they are providing all their details in online
- The Accuracy of prediction can also be improved

#### 1 1.CONCLUSION

The analysis has started from data preprocessing ,handling missing value, exploratory analysis and different models were built like Decision tree model,KNN model,Xgboost model and Random Forest model and there performance were evaluated, as a result the Random Forest model is selected as the best model for predicting the loan approval status of the customer after evaluating its performance, as it got 91% accuracy in prediction. This application is then tested and it functions properly and it also meets all the requirements of the bank in selecting the trustworthy person to provide loan.

#### 1 2.FUTURE SCOPE

In future, payment option can be included in this application for exchanging money between the lender and borrower and bank can verify the customer document online using AI which makes the process of verification simpler and could be made more secure, trustworthy and dynamic weight conformation and in near future this module can be integrated with the module of automated processing system.

### 13.APPENDIX

#### SOURCE CODE

Home.html

```
.center{
   position: absolute;
    top: 58%;
left: 50%;
    width: 100%;
    padding: 0 20px;
text-align: center;
  color: □rgb(0, 20, 31);
    font-weight: 900;
    font-family: 'Gill Sans', 'Gill Sans MT', Calibri, 'Trebuchet MS', sans-serif;
.center .sub_title{
    color: □rgb(5, 21, 65);
    font-size: 18px;
    font-weight: 900;
    font-family:'Gill Sans', 'Gill Sans MT', Calibri, 'Trebuchet MS', sans-serif;
.center .btns{
  margin-top: 20px;
   height: 45px;
width: 150px;
    border-radius: 25px;
    border: none;
    margin: 0 10px;
    border: 1px solid ■white;
```

```
.center .sub_title{
color: □rgb(5, 21, 65);
    font-weight: 900;
    font-family:'Gill Sans', 'Gill Sans MT', Calibri, 'Trebuchet MS', sans-serif;
   margin-top: 20px;
.center .btns button{
   height: 45px;
   width: 150px;
   border-radius: 25px;
   border: none;
   margin: 0 10px;
   border: 1px solid ■white;
   font-weight: 500;
   font-family: fantasy;
   padding: 0 10px;
   color: □rgb(8, 8, 8);
background: ■rgb(249, 250, 249);
.center .btns button:first-child:hover{
   background: ☐rgb(37, 71, 221);
   color: ■rgb(236, 247, 247);
```

#### **Predict**.html

```
C: > Users > ELCOT > Desktop > Ibm project > templates > ◆ predict.html > ♦ html > ♦ style > ≰ .wrapper
       <html lang="en">
<style type="text/css">
          margin: 0;
          padding: 0;
         box-sizing: border-box;
font-family: 'Montserrat', sans-serif;
       body{
   background: □grey;
         padding: 0 10px;
        .wrapper{
max-width: 500px;
         width: 100%;
background: ■#fff;
          margin: 20px auto;
          box-shadow: 4px 4px 25px □#000;
         padding: 30px;
       .wrapper .title{
        font-size: 20px;
         font-weight: 700;
         margin-bottom: 25px;
          text-align: center;
```

```
.wrapper .form{
 width: 100%;
.wrapper .form .inputfield{
  margin-bottom: 15px;
 align-items: center;
.wrapper .form .inputfield label{
 width: 200px;
  color: □#757575;
   margin-right: 10px;
  font-size: 14px;
 width: 100%;
  border: 1px solid ■#d5dbd9;
  padding: 8px 10px;
  border-radius: 3px;
transition: all 0.3s ease;
 width: 100%;
  height: 125px;
 resize: none:
```

```
.wrapper .form .inputfield .textarea{
  width: 100%;
 height: 125px;
.wrapper .form .inputfield .custom_select{
 position: relative;
 width: 100%;
 height: 37px;
content: "";
position: absolute;
 right: 10px;
 border: 8px solid;
 border-color: ■#d5dbd9 transparent transparent transparent;
 -webkit-appearance: none;
 -moz-appearance: none;
 appearance:
  width: 100%;
  height: 100%;
 border: 0px;
  padding: 8px 10px;
```

### Approve .html

```
<!DOCTYPE html>
<html lang="en" dir="ltr">
             <meta charset="utf-8">
               \verb|\ditle>Loan| approva status<| \texttt{title}>
              k rel="stylesheet" href="static/approve.css">
k rel="stylesheet" href="https://cdnjs.cloudflare.com/ajax/libs/font-awesome/5.15.3/css/all.min.css"/>
             <h1>LOAN APPROVAL STATUS</h1>
                \label{eq:h2} $$ \hborder{$\hborder}(h2) = \hborder{$\hborder}(h2) =
                <img src="static/approve.jpg" width="150px" height="150px">
                <h3>Please provide your feedback</h3>
                       <div class="post">
                            <div class="text">Thanks for rating us!</div>
                              <div class="edit">EDIT</div>
                             <input type="radio" name="rate" id="rate-5">
                               <label for="rate-5" class="fas fa-star"></label>
                             <input type="radio" name="rate" id="rate-4"</pre>
                             <label for="rate-4" class="fas fa-star"></label>
<input type="radio" name="rate" id="rate-3">
                               <label for="rate-3" class="fas fa-star"></label>
<input type="radio" name="rate" id="rate-2">
                               <input type="radio" name="rate" id="rate-1">
<label for="rate-1" class="fas fa-star"></label>
```

### reject .html

```
<!DOCTYPE html>
<html lang="en" dir="ltr">
    <meta charset="utf-8">
     <title>Loan approval status</title>

    <h1>LOAN APPROVAL STATUS</h1>
     \label{eq:h2} $$ \hborder{$h2$}{{prediction\_text}}</h2$$
    <img src="static/reject.jpg" width="200px" height="200px">
<h3>Please provide your feedback</h3>
     <div class="container">
       <div class="edit">EDIT</div>

<label for="rate-3" class="fas fa-star"></label>
<input type="radio" name="rate" id="rate-2">
          <label for="rate-2" class="fas fa-star"></label>
<input type="radio" name="rate" id="rate-1">

          <label for="rate-1" class="fas fa-star"></label>
            <header></header>
<div class="textages">
```

## **Model.ipynb**

```
{\tt 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 \ \ Gradient Boosting Classifier, \ Random Forest Classifier
from sklearn.neighbors import KNeighborsClassifier
from sklearn.model_selection import RandomizedSearchCV
import warnings
warnings.filterwarnings('ignore')
{\tt 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
from sklearn.model_selection import cross_val_score
data=pd.read_csv(r"D:\gokulnath\ibm\Loan_Predict.csv")
data.head()
```

```
def decisionTree(x_train,x_test,y_train,y_test):
     dt = DecisionTreeClassifier()
     dt.fit(x_train,y_train)
     pred_test = dt.predict(x_test)
     pred_test
     print('***DecisionTreeClassifier***')
print('Confusion Matrix')
     print(confusion_matrix(y_test,pred_test))
     print('Classification Report')
     print(classification_report(y_test,pred_test))
     print('Score')
     print(dt.score(x_test,y_test))
   decisionTree(x_train,x_test,y_train,y_test)
***DecisionTreeClassifier***
Confusion Matrix
[[48 12]
[14 65]]
Classification Report
              precision
                           recall f1-score support
                   0.77
                              0.80
                                        0.79
                                                     60
```

```
data.shape
   print("\n")
   data.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 614 entries, 0 to 613
Data columns (total 13 columns):
# Column
                   Non-Null Count Dtype
   Loan ID
                    614 non-null
                                   object
                    601 non-null
   Gender
                                   object
2 Married
                    611 non-null
                                   object
3 Dependents
                     599 non-null
                                   object
                    614 non-null
4 Education
                                   object
5 Self_Employed
                     582 non-null
                                    object
6 ApplicantIncome 614 non-null
                                    int64
7 CoapplicantIncome 614 non-null
                                    float64
                     592 non-null
                                    float64
9 Loan Amount Term 600 non-null
                                    float64
10 Credit_History
                     564 non-null
                                    float64
                     614 non-null
11 Property Area
                                   object
```

```
rf = RandomForestClassifier()
        pred_test = rf.predict(x_test)
       print("**** Random Forest Classifier ****")
print('Confusion Matrix')
        print(confusion_matrix(y_test,pred_test))
        print('Classification Report')
        print(classification_report(y_test,pred_test))
        print('Score')
print(rf.score(x_test,y_test))
   randomForest(x_train,x_test,y_train,y_test)
**** Random Forest Classifier ****
Confusion Matrix
[[51 9]
Classification Report
               precision
                           recall f1-score support
                    0.88
                               0.85
                                         0.86
                                                      60
                    0.89
                               0.91
                                         0.90
    accuracy
                                         0.88
   macro avg
                    0.88
                               0.88
                                         0.88
```

```
Knn = KNeighborsClassifier()
       Knn.fit(x_train,y_train)
       pred_test = Knn.predict(x_test)
       print("**** KNeigborsClassifier ****")
       print('Confusion Matrix')
       print(confusion_matrix(y_test,pred_test))
       print('Classification Report')
       print(classification_report(y_test,pred_test))
       print('Score')
       print(Knn.score(x_test,y_test))
   KNN(x_train,x_test,y_train,y_test)
**** KNeigborsClassifier ****
Confusion Matrix
[[38 22]
[27 52]]
Classification Report
             precision
                          recall f1-score
                  0.58
                             0.63
                                       0.61
                                                   60
                  0.70
                             0.66
                                       0.68
```

GITHUB: https://github.com/IBM-EPBL/IBM-Project-11641-1664170589

**DEMO LINK:**