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### 1. INTRODUCTION

## 1.1Project Overview

Recently, clinicians have been actively engaged in improving medical diagnoses. The use of artificial intelligence and machine learning in combination with clinical findings has further improved disease detection. In the modern era, with the advantage of computers and technologies, one can collect data and visualize many hidden outcomes such as dealing with missing data in medical research. Statistical machine learning algorithms based on specific problems can assist one to make decisions. Machine learning (ML), data-driven algorithms can be utilized to validate existing methods and help researchers to make potential new decisions.

### 1.2Purpose

People who want to know whether they have liver disease or not.

### 2. LITERATURE SURVEY

### 2.1 References

S. No	Paper	Year	Citation	Methodologies used
1	Statistical Machine Learning Approaches to Liver Disease Prediction	2021	Mostafa, F.; Hasan, E.; Williamson, M.; Khan, H. Statistical Machine Learning Approaches to Liver Disease Prediction. Livers 2021, 1, 294–312. https://doi.org/10.3390/ livers1040023	The purpose of this study was to extract significant predictors for liver disease from the medical analysis of 615 humans using ML algorithms. Data visualizations were implemented to reveal significant findings such as missing values.

			Ι	
2	Performance Analysis of Liver Disease Prediction Using Machine Learning Algorithms	2018	International Research Journal of Engineering and Technology (IRJET) www.irjet.net	P.Rajeswari,G.Sophia Reena et al.,[2010]has proposed the data classification is based on liver disorder. The training dataset is developed by collecting data from UCI repository consists of 345 instances with 7 different attributes. This paper deals with results in the field of data classification obtained with Naïve Bayes algorithms .FT tree algorithms, and KStar algorithms and on the whole performance made know FT Tree algorithm when tested on liver disease datasets, time taken to run the data for result is fast when compare to other algorithm with accuracy of 97.10%Based on the experimental results the classification accuracy is found to be better using FT Tree algorithm compare to
				better using FT Tree

3	Liver Disease Prediction System using Machine Learning Techniques	2021	International Journal of Engineering Research & Technology (IJERT)  Vol. 10 Issue 06, June-2021	Researchers are attempting to discover models for early diagnosis of illness utilizing biomedical information. Since the most recent couple of decades, they have utilized a parcel of models for early finding, each with their very own advantages and disadvantages. In this research, a CHIRP based model is proposed for the early forecast of liver disease.
4	Liver Disease Prediction Using Machine Learning Algorithm	2021	Authors:  Sambit Mohanty, Pradosh Kumar Gantayat, Sachikanta Dash, Bhabani P. Mishra & Shiba Ch. Barik  Conference paper  First Online: 05 May 2021  Part of the Advances in Intelligent Systems and Computing book series (AISC,volume 1407)	We can use these data to improve our healthcare services or proper identification of diseases. We have collected patient data from open source platform and applied various kinds of data analysis techniques and machine learning (ML) approaches applied to see the pattern of the

				data sets. Then, a performance comparison between these models is made to get highly accurate model for predicting liver disease.
5	Liver disease prediction using machine learning and deep learning: A comparative study	2022	March 2022     Intelligent Decision     Technologies 16(3):1- 14  DOI:10.3233/IDT-210065  Authors: Bhawna Singla Soham Taneja Rishika Garg Preeti Nagrath	This research aims to impart insight additional to the current state-of-the-art discoveries by focusing on a comparative analysis of some of the best ML/DL techniques which haven't been scrutinized altogether yet.

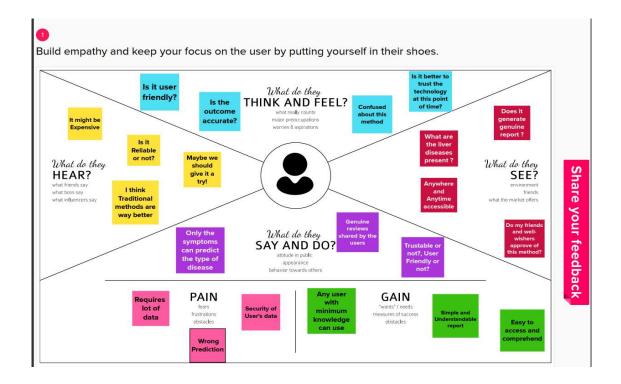
#### 2.2 Problem Statement Definition

Chronic liver diseases include chronic hepatitis, fibrosis, and cirrhosis. Hepatitis can occur from viral infection (e.g., hepatitis c virus) or autoimmune origin. When liver disease is diagnosed at an earlier stage, in between infection and fibrosis but before cirrhosis, liver failure can be avoided. The use of artificial intelligence and machine learning in combination with clinical findings has further improved disease detection. Statistical machine learning algorithms based on specific problems can assist one to make decisions.

### 3.IDEATION & PROPOSED SOLUTION

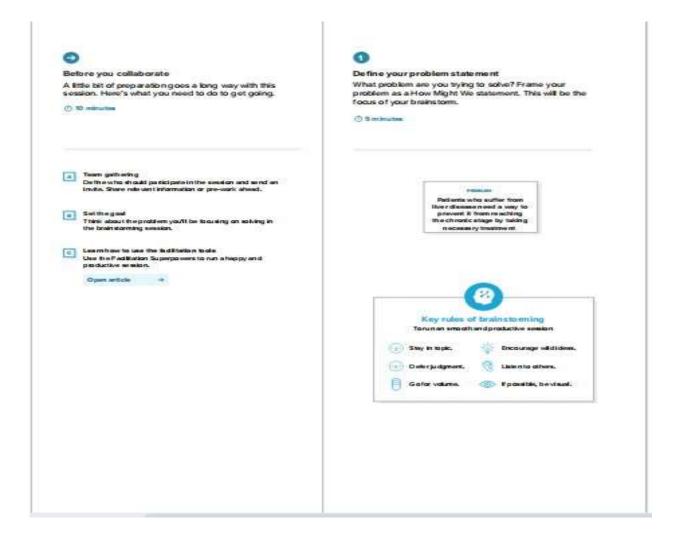
### 3.1 Empathy Map Canvas

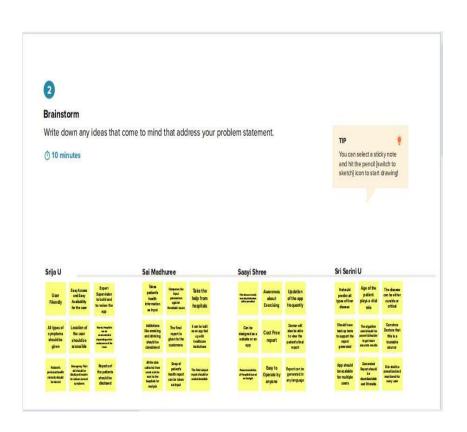
An empathy map is a simple, easy-to-digest visual that captures knowledge about a user's behaviours and attitudes. It is a useful tool to help teams better understand their users. Creating an effective solution requires understanding the true problem and the person who is experiencing it. The exercise of creating the map helps participants consider things from the user's perspective along with his or her goals and challenges.

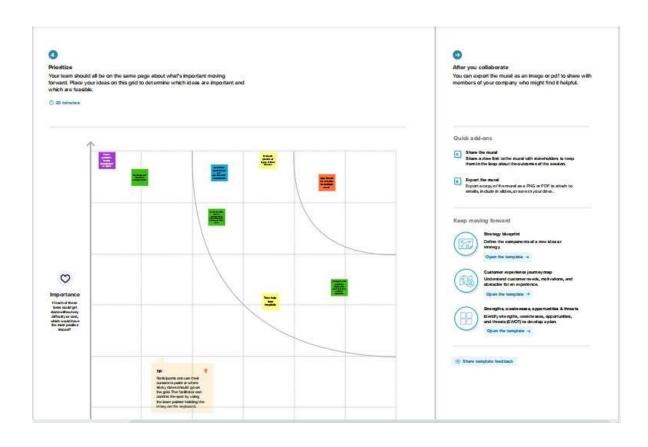


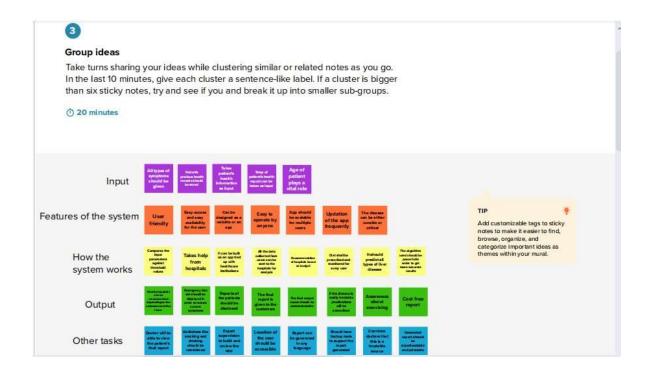
# 3.2 Ideation, Brainstorming & Prioritize

# Brainstorming









# 3.3 Proposed Solution

### Project Design Phase-I Proposed Solution Template

Date	26 September 2022
Team ID	PNT2022TMID53422
Project Name	Statistical Machine Learning Approaches to Liver Disease Prediction
Maximum Marks	2 Marks

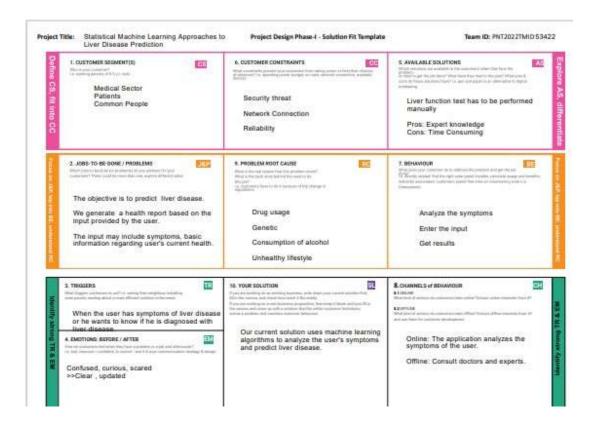
#### Proposed Solution Template:

Project team shall fill the following information in proposed solution template:

S.No	Parameter	Description
L	Problem Statement (Problem to be solved)	To detect disease, healthcare professionals need to collect samples from patients which can east both time and money. Often, more than one kind of test or many samples are needed from the patient to accumulate all the necessary information for a better diagnosis. There is a need to find better ways to detect and diagnose liver disease with more accuracy.
2	idea/ Solution description	Statistical machine learning algorithms based on specific problems can assist one to make decisions. Machine learning (ML), data-driven algorithms can be utilized to validate existing methods and help researchers to make potential new decisions
3.	Novetty / Uniqueness	Various kinds of data sets, such as blood panels with liver function tests, histologically stained slide images, and the presence of specific molecular markers in blood or tissue samples, have been used to train classifier algorithms to predict liver disease with good accuracy.
4.	Social Impact / Customer Satisfaction	Application of the ML methods can help reduce the total burden of liver disease on public health worldwide by improving recognition of risk factors and diagnostic variables. More importantly, for chronic liver disease, detecting liver disease at earlier stages or in hidden cases by ML.

		could decrease liver-related mortality, transplants, and/or hospitalizations.
5.	Business Model (Revenue Model)	The global liver disease diagnostics market size was valued at USD 29.3 billion in 2019 and is estimated to grow at a compound annual growth rate (CAGR) of 6.5% from 2020 to 2027. Rising prevalence of scute and chronic liver diseases is one of the major factors expected to drive the market for liver disease diagnostics
6	Scalability of the Solution	The described ML methods can assist health sectors to achieve a better diagnosis providing effective results in identifying groups or levels within medical data to facilitate healthcare workers. The machine learning algorithms presented in this study can support medical experts but are not the alternative when making decisions from ML classifiers for diagnostic pathways.

### 3.4 Problem Solution fit



# 4. REQUIREMENT ANALYSIS

# 4.1 Functional requirement

FR No.	Functional Requirement (Epic)	Sub Requirement (Story / Sub-Task)
FR-1	User Registration	Registration through
		Gmail
		Registration through
		Healthcare Portals
FR-2	User Confirmation	Confirmation via Email
		Confirmation via OTP
FR-3	User Application	Filling of application
		Modification of application
		Verification of application
FR-4	User Verification	Verifying through Credentials through Database
FR-5	Database Update	1.Updating Data in
	·	Database
		2.Altering data in
		Database
		3.Deleting Data in
		Database
FR-6	Activities Tracking	Track the activities and keep the database updated

# 4.2 Non-Functional requirements

FR No.	Non-Functional Requirement	Description
NFR-1	Usability	The prediction model can be used by users, doctors and health care institutions
NFR-2	Security	The prediction model is not vulnerable to any brute force attack or any security attacks.
NFR-3	Reliability	The prediction model can only be accessed by the users who possess the username and password.

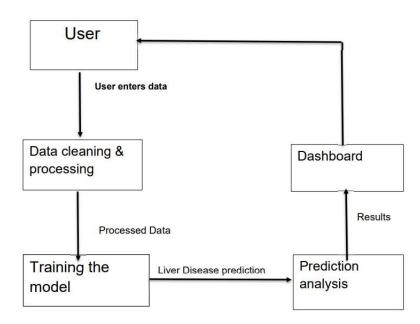
NFR-4	Performance	The prediction model can take in huge amounts of data, scalable for large-scale users and can do prediction as well as visualization.

NFR-5	Availability	The prediction model can be accessed at anytime and anywhere.
NFR-6	Scalability	The prediction model can be used by multiple users and multiple hospitals at the same time

### **5.PROJECT DESIGN**

## **5.1 Data Flow Diagrams**

A Data Flow Diagram (DFD) is a traditional visual representation of the information flows within a system. A neat and clear DFD can depict the right amount of the system requirement graphically. It shows how data enters and leaves the system, what changes the information, and where data is stored.



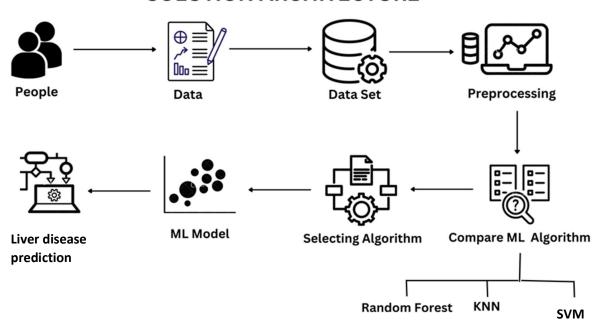
### **5.2 Solution & Technical Architecture**

#### **Solution Architecture:**

Solution architecture is a complex process – with many sub-processes – that bridges the gap between business problems and technology solutions. Its goals are to:

- Find the best tech solution to solve existing business problems
- Describe the structure, characteristics, behaviour, and other aspects of the software to project stakeholders
- Define features, development phases, and solution requirements
- Provide specifications according to which the solution is defined, managed, and delivered.

### **SOLUTION ARCHITECTURE**



# **5.3** User Stories

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	Registration	USN-1	As a user, I can register for the application by entering my email, password, and confirming password.	5	High	Srija
Sprint-1		USN-2	As a user, I will receive confirmation email once I have registered for the application	5	High	Sri Sarini
Sprint-1	Login	USN-3	As a user, I can log into the application by entering email & password	10	Medium	Sai Madhuree
Sprint-2	Input Necessary Details	USN-4	As a user, I can give input test details to predict liver disease	15	High	Saayi Shree
Sprint-2	Data Pre-Processing	USN-5	Transform raw data into appropriate format for prediction	5	High	Srija
Sprint-3	Prediction of Liver Disease	USN-6	As a user I can get the results of liver disease prediction	15	High	Sri Sarini
Sprint-3		USN-7	As a user I can get accurate results of liver disease	5	Medium	Sai Madhuree
Sprint-4	Review	USN-8	As an admin I reinforce the result of prediction	20	High	Saayi Shree

## 6. PROJECT PLANNING & SCHEDULING

# **6.1 Sprint Planning & Estimation**

Product Backlog, Sprint Schedule, and Estimation

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	Registration	USN-1	As a user, I can register for the application by entering my email, password, and confirming password.	5	High	Srija
Sprint-1		USN-2	As a user, I will receive confirmation email once I have registered for the application	5	High	Sri Sarini
Sprint-1	Login	USN-3	As a user, I can log into the application by entering email & password	10	Medium	Sai Madhuree
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Sprint-3		USN-7	As a user I can get accurate results of liver disease	5	Medium	Sai Madhuree
Sprint-4	Review	USN-8	As an admin I reinforce the result of prediction	20	High	Saayi Shree

Sprint	Total Story Points	Duration	Sprint Start Date	Sprint End Date (Planned)	Story Points Completed (as on Planned End Date)	Sprint Release Date (Actual)
Sprint-1	20	6 Days	24 Oct 2022	29 Oct 2022	20	29 Oct 2022
Sprint-2	20	6 Days	31 Oct 2022	05 Nov 2022		
Sprint-3	20	6 Days	07 Nov 2022	12 Nov 2022		
Sprint-4	20	6 Days	14 Nov 2022	19 Nov 2022		

#### Velocity

Imagine we have a 10-day sprint duration, and the velocity of the team is 20 (points per sprint). Let's calculate the team's average velocity (AV) per iteration unit (story points per day)

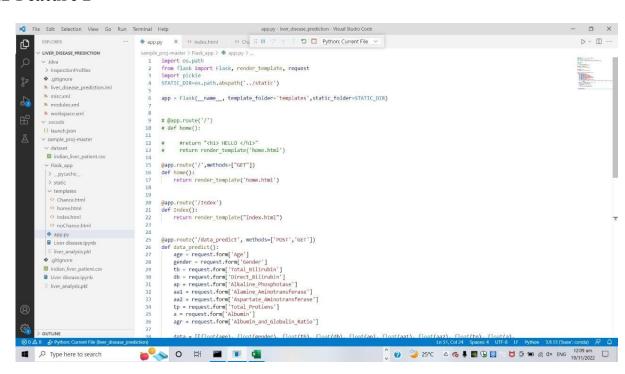
$$AV = \frac{sprint\ duration}{velocity} = \frac{20}{10} = 2$$

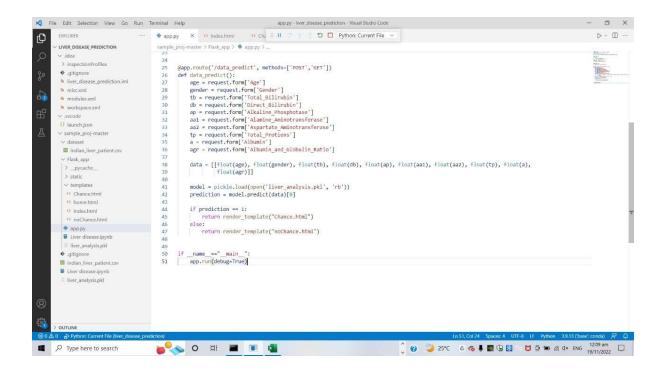
### 7. CODING & SOLUTIONING

### 7.1 Feature 1

Liver Diseas	se Prediction		
oge:  7.3  Ukaline Phosphotase:  490  uspartate Aminotransferase:  33  ubumin:  3.3	Gender:  1 Direct Bilirubin:  4.1 Alamine Aminotransferase: 60 Total Protiens: 7 Albumin and Globulin Ratio: 0.88		

### 7.2 Feature 2









#### 8.TESTING

### **8.1 Test Cases**

Tes t cas e ID	Feature Type	Compon ent	Test Scenari o	Pre- Requisi te	Steps To Execute	Tes t Dat a	Expected Result	Actual Result	Stat us	Comme nts	TC for Automation( Y/N)	BU G ID	Execut ed By
tc0 1	Functio nal	Page	Verify user is able to click on Predict button		1.Enter URL and fill the form 2.Click on Predict button		Loan form should display	Worki ng as expect ed	Pass				
tc0 2	Functio nal	Home Page	The web page is getting refresh ed		1.Autom atic page reload		Loan form must appear automatic ally after page reload	Worki ng as expect ed	Fail	No steps needed	Υ	BU G- 123 4	
tc0 3	Functio nal	Home page	Field address validati on		1. Double- click on the E- mail address field		User should navigate to E-mail address field	Worki ng as expect ed	Pass				
tc0	Functio nal	Output page	Loan Credibil ity predict ed output		1. Click on predict button 2. View the predicted results		User should access the Loan credibility predicted result	Worki ng as expect ed	Pass				

# **8.2** User Acceptance Testing

# 1. Purpose of Document

The purpose of this document is to briefly explain the test coverage and open issues of the [ProductName] project at the time of the release to User Acceptance Testing (UAT).

# 2. Defect Analysis

This report shows the number of resolved or closed bugs at each severity level, and how they were resolved

Resolution	Severity 1	Severity 2	Severity 3	Severity 4	Subtotal
------------	---------------	---------------	---------------	---------------	----------

By Design	9	3	1	2	15
Duplicate	0	0	4	0	4
External	1	2	0	0	3
Fixed	10	5	4	21	40
Not Reproduced	0	0	0	0	0
Skipped	0	0	0	0	0
Won't Fix	0	0	0	0	0
Totals	20	9	9	23	64

# 1. Test Case Analysis

This report shows the number of test cases that have passed, failed, and untested

Section	Total Cases	Not Tested	Fail	Pass
Print Engine	5	0	0	5
Client Application	46	0	0	46
Security	4	0	0	4
Outsource Shipping	3	0	0	3
Exception Reporting	7	0	0	7
Final Report Output	4	0	0	4
Version Control	2	0	0	2

### 9. RESULTS

### 9.1 Performance Metrics

## **Model Performance Testing:**

Project team shall fill the following information in the model performance testing template.

S.No.	Parameter	Screenshot / Values
1.	Data Responsiveness	The ML model takes about 0.3 seconds to process the dataset. The credibility result is predicted in approximately 0.9 seconds.
2.	Utilisation of Data Filters	Sufficient data filters have been used for ideal model building
3.	Effective User Story	No of Scene Added - 15
4.	Descriptive Reports	No of Visualisations / Graphs - 13

### 10. ADVANTAGES & DISADVANTAGES

### Advantage:

# Anyone can use the application.

Our application can be used by anyone with minimum knowledge. The Usability of the App is one of the advantage.

# User Friendly UI

User Friendly User Interface enables the users to easily understand how the application works. Therefore, Encouraging users to use the application whenever they want.

### Remote Access

At home diagnosis in case of emergencies as the results are much easier and faster to get. Simple diagnosis can be done at home without the need to go to the hospital.

### Disadvantage:

## Does not replace hospitals

The user cannot completely rely on the application and this application cannot replace hospitals or prevent the users to go to the hospitals. It is just an emergency or a support application for the users.

### Prediction may not be accurate

The user cannot completely trust the diagnosis done through the application. The user may need to double check by visiting a doctor. But this may not happen everytime. This is advised just to be on a safer side.

### 11. CONCLUSION

The analysis starts from data cleaning and processing missing value, exploratory analysis and finally model building and evaluation of the model. The best accuracy on public test set is when we get higher accuracy score and other performance metrics which will be found out. This project can help to predict the liver disease of the user.

### 12. FUTURE SCOPE

The health or life of the user is relied on this application therefore the accuracy can be improved in order to completely trust the application. This application can also help healthcare workers and can serve as a support system.

### 13. APPENDIX

**Source Code** 

app.py

import os.path

from flask import Flask, render\_template, request import pickle STATIC DIR=os.path.abspath('../static')

```
app = Flask( name , template folder='templates', static folder=STATIC DIR)
# @app.route('/')
# def home():
    #return "<h1> HELLO </h1>"
#
#
    return render template('home.html')
@app.route('/',methods=["GET"])
def home():
  return render template('home.html')
@app.route('/Index')
def Index():
  return render template("Index.html")
@app.route('/data predict', methods=['POST','GET'])
def data predict():
  age = request.form['Age']
  gender = request.form['Gender']
  tb = request.form['Total_Bilirubin']
  db = request.form['Direct Bilirubin']
  ap = request.form['Alkaline Phosphotase']
 aa1 = request.form['Alamine Aminotransferase']
  aa2 = request.form['Aspartate Aminotransferase']
  tp = request.form['Total Protiens']
  a = request.form['Albumin']
  agr = request.form['Albumin_and_Globulin_Ratio']
  data = [[float(age), float(gender), float(tb), float(db), float(ap), float(aa1),
   float(aa2), float(tp), float(a),
        float(agr)]]
  model = pickle.load(open('liver analysis.pkl', 'rb'))
  prediction = model.predict(data)[0]
  if prediction == 1:
    return render template("Chance.html")
  else:
    return render template("noChance.html")
```

```
if name=="main "
app.run(debug=True)
index.html
<form action="/predict" method="post">
   <br>
   <br>
   <br>
        Age:  
     
        
;        
p; 
        
; 
          
   <br>
   <br>
   <input type="number" id="Age" placeholder="Age " name="Age"</pre>
style="width: 190px; margin-left:17px;"> <input type="number" id="Gender"
     placeholder="Gender (Male:1, female:0)" name="Gender"
style="width: 190px; margin-left: 97px;">
   <br>
   <br>
         Total Bilirubin:    
      
       
        
;       Direct Bilirubin:
   <br/><br><input type="text" id= "Total Bilirubin"
placeholder="Total Bilirubin"
     name="Total Bilirubin" style="width: 190px; margin-left:17px;">
    <input type="text" id="Direct Bilirubin" placeholder="Direct</pre>
Bilirubin"
     name="Direct Bilirubin" style="width: 190px; margin-left: 97px;">
   <br>
   <br>
```

```
        
;       Alamine Aminotransferase:
    <br>
    <br>
    <input type="text" id="Alkaline Phosphotase" placeholder="Alkaline</pre>
Phosphotase"
      name="Alkaline Phosphotase" style="width: 190px; margin-left:
17px;">
     <input type="text" id="Alamine Aminotransferase"</pre>
      placeholder="Alamine Aminotransferase"
name="Alamine Aminotransferase" style="width: 190px; margin-left: 95px;">
    <br>
    <br>
               
       
      Total Protiens:
    <br>
    <br>
     <input type=" text" id="Aspartate_Aminotransferase"</pre>
      placeholder="Aspartate_Aminotransferase"
name="Aspartate Aminotransferase" style="width: 190px; margin-left:
17px;">
    <input type=" text" id="Total Protiens" placeholder="Total Protiens"</pre>
name="Total Protiens" style="width: 190px; margin-left: 97px;">
    <br>
    <br>
             
     
        
;        
        
;       Albumin and Globulin
Ratio:
    <br>
    <br>
    <input type="text" id="Albumin" placeholder=" Albumin " name="
Albumin" style="width: 190px; margin-left:17px;">
     <input type="text" id="Albumin and Globulin Ratio"</pre>
```

# 13.2 GitHub& Project:

https://github.com/IBM-EPBL/IBM-Project-18196-1659680664/upload/main/Project%20Development%20Phase