# UNIVERSITY ADMIT ELIGIBILITY PREDICTOR

### **IBM PNT2022TMID02243**

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### 1. INTRODUCTION

### 1.1 Project Overview

Students frequently have concerns over their opportunities to enrol in university programmes. This project's goal is to assist students in narrowing down their choices of institutions by improving their profiles. The outcome that was projected provides them with a reasonable estimate of their prospects of being accepted into a certain university. Students who are presently preparing or who will be prepared could also benefit from this analysis because it should assist them to obtain a better picture.

### 1.2 Purpose

Students will be able to check if they are eligible to enrol in university programmes or not.

### 2. LITERATURE SURVEY

### 2.1 References

S. No	Paper	Year	Citation	Methodologies used
1	Graduate Admission Chance Prediction Using Deep Neural Network	2020	2020 IEEE International Women in Engineering (WIE) Conference on Electrical and Computer Engineering (WIECON-ECE)	The DNN model has been compared with existing methods in terms of different performance metrics including mean squared error (MSE), root mean squared error (RMSE), mean absolute error (MAE), R-squared score. It has shown the most promising result that includes an R-squared score of 0.8538 and MSE of 0.0031.
2	A Statistical Approach to Graduate Admissions' Chance Prediction	2020	Navoneel Chakrabarty, Siddhartha Chowdhury, Srinibas Rana, International Research Journal of Engineering and Technology (IRJET) e-ISSN: 2395-0056, Volume: 09 Issue: March 2020	The developed models are gradient boosting regressor and linear regression model. Gradient boosting regressor have to score of 0.84. That surpassing the performance of linear regression model. They computed different other performance error metrics like mean absolute error, mean square error, and root mean square error.

3	Literature Survey on Student's Performance Prediction in Education using Data Mining Techniques	2017	Mukesh Kumar, Disha Handa,International Journal of Education and Management Engineering 6(6):40-49  DOI:10.5815/ijme.2017.06.05	Article aims to provide a great knowledge and understanding of different data mining techniques which have been used to predict the student progress and performance and hence how these prediction techniques help to find the most important student attribute for prediction. It wants to improve the performance of the student in academic by using best data mining techniques
4	Prediction for University Admission using Machine Learning	2020	Chithra Apoorva D A, Malepati ChanduNath, Peta Rohith, Bindu Shree.S, Swaroop.S, International Journal of Recent Technology and Engineering (IJRTE) ISSN: 2277-3878 (Online), Volume-8 Issue-6, March 2020	Problem Understanding, Data Understanding, Data Preparation, Building models, Evaluation

5 Approach Predicting Enrollmen University	for Student t in the	2021 XXX International Scientific Conference Electronics (ET), Date of Conference: 15-17 September 2021Electronic ISBN:978-1-6654-4518-4, Print on Demand(PoD) ISBN:978-1-6654-4519-1	The RepTree algorithm performed the best with a True Positive (TP) rate of 0.902 and a False Positive (FP) rate of 0.148. The algorithm with the lowest performance was NaiveMulti with a TP rate of only 0.790. However, the range between the worst and the best-performing algorithms was 14.18%.
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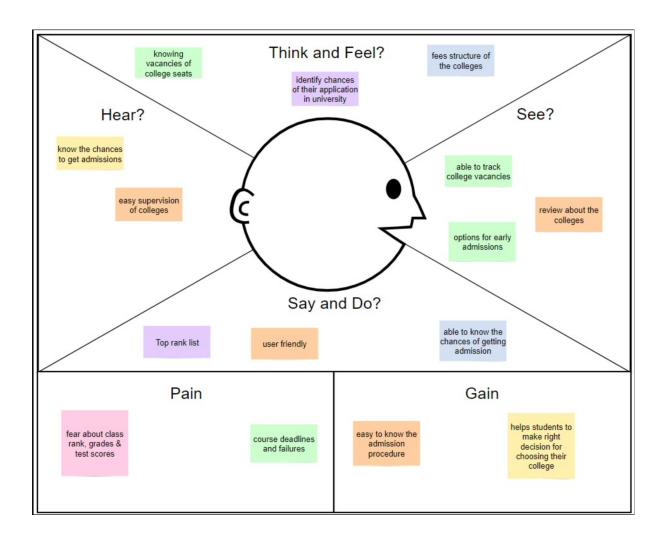
### 2.2 Problem Statement Definition

Students frequently have concerns over their opportunities to enrol in university programmes. This project's goal is to assist students in narrowing down their choices of institutions by improving their profiles. The outcome that was projected provides them with a reasonable estimate of their prospects of being accepted into a certain university. Students who are presently preparing or who will be prepared could also benefit from this analysis because it should assist them to obtain a better picture.

### 3. IDEATION & PROPOSED SOLUTION

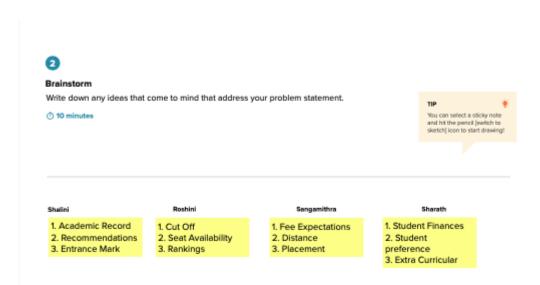
### 3.1 Empathy Map Canvas

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



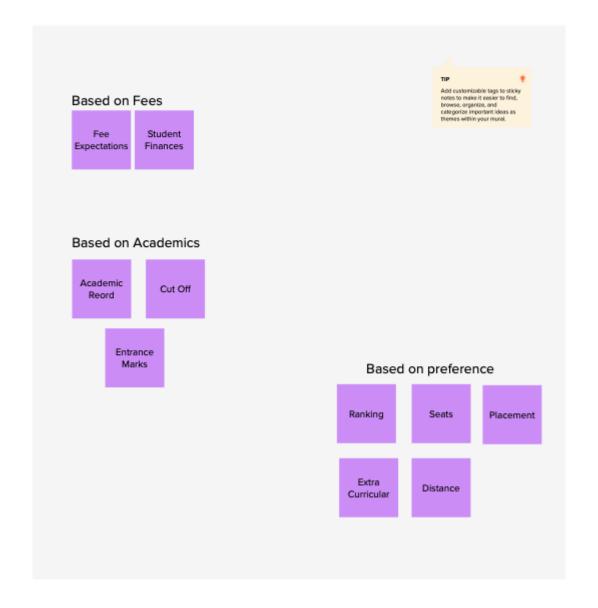




### Group ideas

Take turns sharing your ideas while clustering similar or related notes as you go. In the last 10 minutes, give each cluster a sentence-like label. If a cluster is bigger than six sticky notes, try and see if you and break it up into smaller sub-groups.

① 20 minutes

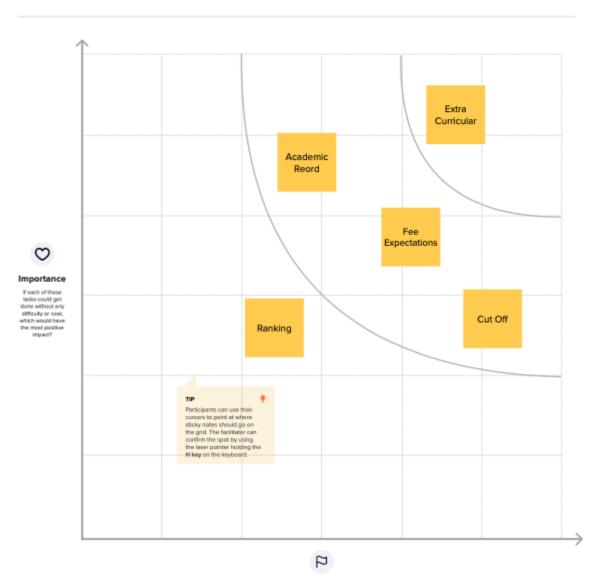




### Prioritize

Your team should all be on the same page about what's important moving forward. Place your ideas on this grid to determine which ideas are important and which are feasible.

① 20 minutes



### Feasibility

feasible than others? (Cost, time, effort, complexity, etc.



### After you collaborate

You can export the mural as an image or pdf to share with members of your company who might find it helpful.

### Quick add-ons



A Share the mural

Share a view link to the mural with stakeholders to keep them in the loop about the outcomes of the session.

Export the mural
 Export a copy of the mural as a PNG or PDF to attach to emails, include in sides, or save in your drive.

### Keep moving forward



Define the components of a new idea or strategy.

Open the template ->



### Customer experience journey map

Understand customer needs, motivations, and obstacles for an experience.

Open the template  $\rightarrow$ 



### Strengths, weaknesses, opportunities & threats

Identify strengths, weaknesses, opportunities, and threats (SWOT) to develop a plan.

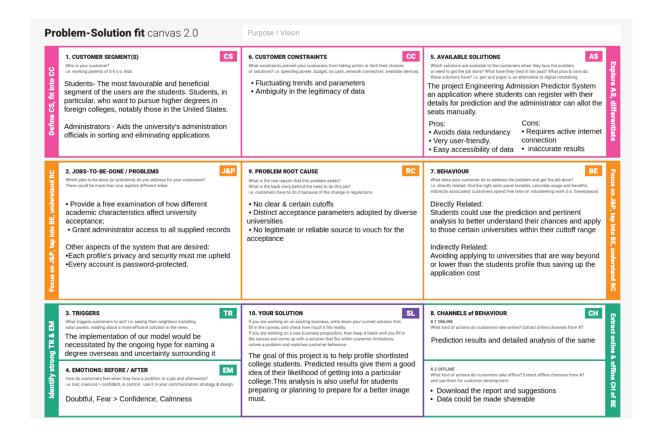
Open the template +

Share template feedback

# 3.3 Proposed Solution

S.No.	Parameter	Description
1	Problem Statement (Problem to be solved)	Many students worry and wonder if they will get into the university they desire. In rare cases, students are not able to attend the universities of their choice. They won't have time to submit applications to other colleges at the last minute. The student feels a lot of stress as a result of this. It would be preferable if students could find out in advance if they can get into the university of their choice so that they can be ready for the worst-case scenario.
2.	Idea / Solution description	A program based on machine learning algorithm to classify the result with the help of the given details like academic qualifications, GRE scores, TOEFL scores etc.
3.	Novelty / Uniqueness	High accurate machine learning algorithm without compromising the features and enhanced UI
.4.	Social Impact / Customer Satisfaction	Really helpful and more accurate, as judged by customer feedback and prediction accuracy of more than 90%
5.	Business Model (Revenue Model)	The goals of the research were initially determined in the process of spending a significant amount of time comprehending the problem statement by analysing the concerns of students over the current application procedure thereby turning this into a profitable business model concerning students.
6.	Scalability of the Solution	The application is extremely user-friendly and accessible from anywhere in the globe without the need to register or deal with any additional hassles. It is open-sourced, and there are no restrictions for specific users.

### 3.4 Problem Solution fit



# 4.1 Functional requirement

FR No.	Functional Requirement (Epic)	Sub Requirement (Story / Sub-Task)
FR-1	User Details	Submit the Documents:      GRE or/and TOEFL score sheet     Curriculum Vitae (CV)     Statement of Purpose (SoP)     Letter of Recommendation (LOR)
FR-2	User Requirements	<ul> <li>Relevant documents and details are uploaded in the appropriate location on the website.</li> <li>Based on the uploads, the system would collect all the necessary information.</li> <li>The candidate's list of all possible universities would be displayed based on the collected information.</li> </ul>

### **4.2 Non-Functional requirements**

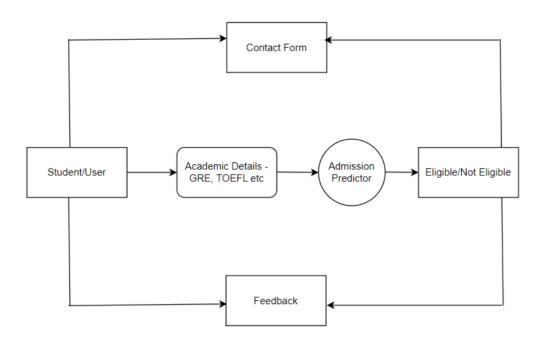
FR No.	Non-Functional Requirement	Description
NFR-1	Usability	<ul> <li>Easy to use and self-explaining website.</li> <li>Easy navigation between pages.</li> <li>Simple structure for faster access.</li> </ul>
NFR-2	Reliability	• Ensembling the outputs of various ML models.
NFR-3	Performance	<ul> <li>Web Based Application.</li> <li>Ability to indicate user inputs of erroneous data types.</li> </ul>

NFR-4	Availability	<ul> <li>Application is available 24 / 7 as it is hosted on IBM cloud.</li> <li>Simple web browser is enough to access the website.</li> </ul>
NFR-5	Scalability	<ul> <li>Can be extended for other types of loans.</li> <li>Aadhar and PAN verification can also be implemented.</li> </ul>

### 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.

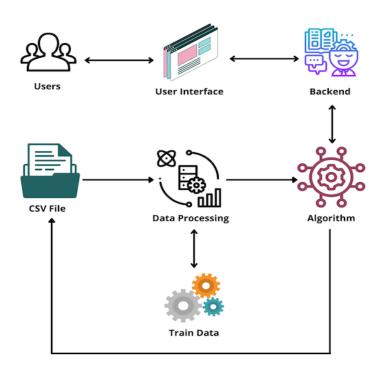


### 5.2 Solution & Technical Architecture

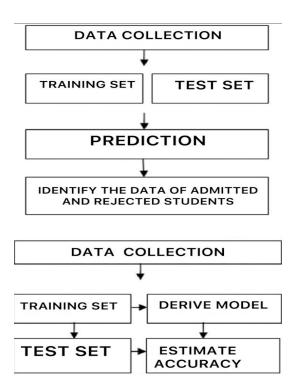
### **Solution Architecture:**

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, behavior, 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.



### **Technical Architecture**:



### **5.3 User Stories**

User Type	Functional  Requiremen t (Epic)	User Story Numb er	User Story / Task	Acceptanc e criteria	Priority	Release
Student	Dashboard	USN-1	User can view and enter details in the website.	I can access the university landing page	Medium	Sprint 1
		USN-2	Queries can be filled, and feedback can be given.	I can fill and submit the contact form	Low	Sprint 2
	Admission	USN-3	Admission eligibility can be predicted for universities	I can get result as either eligible/not eligible	High	Sprint 1
	Updates	USN-4	Updates can be given	I can view the updates	Low	Sprint 3
	Feedback	USN-5	As a user, I can give a positive or negative feedback	can give either good or bad feedback	Low	Sprint 4

### 6. PROJECT PLANNING & SCHEDULING

# **6.1 Sprint Planning & Estimation**

Product Backlog, Sprint Schedule, and Estimation

Sprint	Functional Requireme nt (Epic)	User Stor y Num ber	User Story / Task	Story Poi nts	Priority	Team Members
Sprint-1	Registration	USN-1	As a user, I can register for the application by entering my email, password, and confirming my password	10	High	Roshini V Sangamithra V Shalini M Sharath Srinivas D
Sprint-2	Login	USN-3	Visualise the dataset	5	Low	Roshini V Sangamithra V Shalini M Sharath Srinivas D
Sprint-2	Shortlist	USN-4	Pre-process the dataset	10	Low	Roshini V Sangamithra V Shalini M Sharath Srinivas D
Sprint-3	Predict	USN-5	KNN model building	20	High	Roshini V Sangamithra V Shalini M Sharath Srinivas D
Sprint-4	Preference	USN-6	Decision Tree model building	20	High	Roshini V Sangamithra V Shalini M Sharath Srinivas D

### **6.2 Sprint Delivery Schedule**

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	15	6 Days	24 Oct 2022	29 Oct 2022	15	29 Oct 2022
Sprint-2	15	6 Days	31 Oct 2022	05 Nov 2022	15	05 Nov 2022
Sprint-3	10	6 Days	07 Nov 2022	12 Nov 2022	10	12 Nov 2022
Sprint-4	20	6 Days	14 Nov 2022	19 Nov 2022	20	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 = 60 story points/24 days = 2.5

### **Burndown velocity:**

A burndown chart is a graphical representation of work left to do versus time. It is often used in agile software development methodologies such as Scrum. However, burn down charts can be applied to any project containing measurable progress over time.



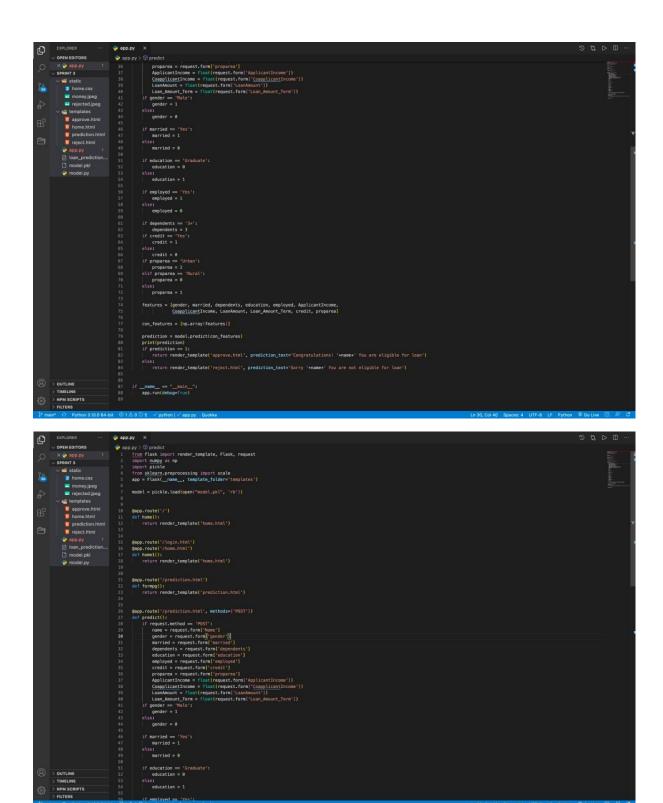


### 7. CODING & SOLUTIONING

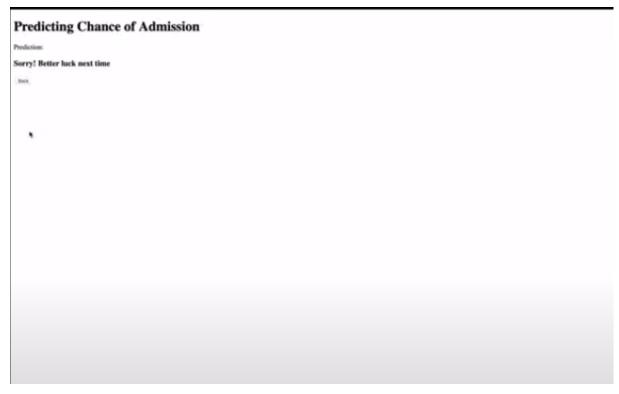
### 7.1 Feature 1

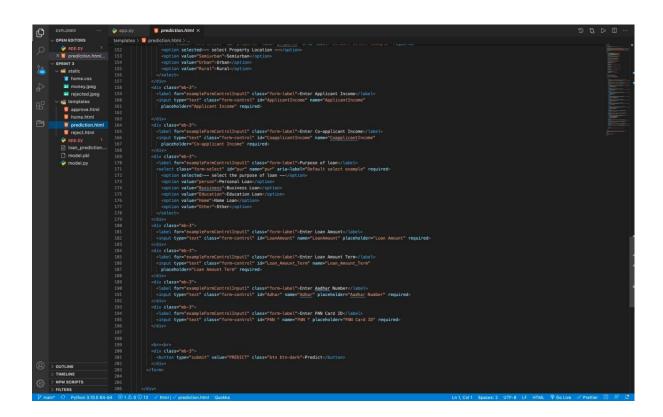


# Predictions: Congrated You're elligible The A.



### **7.2 Feature 2**





### 7.3 Database Schema

The database used here in this project was Admission\_Predict.csv. The sample screenshot of the database are

w	rch this file								
	Serial No.	GRE Score	TOEFL Score	University Rating	SOP	LOR	CGPA	Research	Chance of Admit
2	1	337	118	4	4.5	4.5	9.65	1	0.92
3	2	324	107	4	4	4.5	8.87	1	0.76
4	3	316	104	3	3	3.5	8	1	0.72
5	4	322	110	3	3.5	2.5	8.67	1	0.8
6	5	314	103	2	2	3	8.21	0	0.65
7	6	330	115	5	4.5	3	934	1	0.9
8	7	321	109	3	3	4.	8.2	1	0.75
9	8	308	101	2	3	4	7.9	0	0.68
8	9	302	102	1	2	1.5	8	0	0.5
1	10	323	108	3	3.5	3	8.6	0	0.45
2	11	325	106	3	3.5	4	8.4	1	0.52
3	12	327	111	4	4	4.5	9	1	0.84
4	13	328	112	4	4	45	9.1	1	0.78
5	14	307	109	3	4	3	8	1	0.62
6	15	311	104	3	3.5	2	8.2	1	0.61
7	16	314	105	3	3.5	2.5	8.3	0	0.54
8	17	317	107	3	4	3	8.7	0	0.66
9	18	319	106	3	4	3	8	1	0.65
8	19	318	110	3	4	3	8.8	0	0.63
1	20	303	102	3	3.5	3	8.5	0	0.62
2	21	312	107	3	3	2	7.9	1	0.64
3	22	325	114	4	3	2	8.4	0	0.7
4	23	328	116	5	5	5	9.5	1	0.94
5	24	334	119	5	5	4.5	9.7	1	0.95
6	25	336	119	5	4	3.5	9.8	1	0.97

### 8. TESTING

### **8.1 Test Cases**

Section	Total Cases	Not Tested	Fail	Pass
Print Engine	7	0	0	7
Client Application	51	0	0	51
Security	2	0	0	2
Outsource Shipping	4	0	0	3
Exception Reporting	9	0	0	9
Final Report Output	4	0	0	4
Version Control	2	0	0	2

### **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 University Admit Eligibility Predictor 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	10	4	2	3	19
Duplicate	0	0	0	0	0
External	2	3	0	1	6
Fixed	11	2	4	20	37
Not Reproduced	0	0	0	0	0
Skipped	0	0	1	1	2
Won't Fix	0	0	0	0	0
Totals	24	14	13	26	64

### 1. Test Case Analysis

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

Section	<b>Total Cases</b>	Not Tested	Fail	Pass
Print Engine	7	0	0	7
Client Application	51	0	0	51
Security	2	0	0	2
Outsource Shipping	3	0	0	3
Exception Reporting	9	0	0	9
Final Report Output	4	0	0	4
Version Control	2	0	0	2

### 9. RESULTS

### 9.1 Performance Metrics

### **Model Performance Testing:**

Measure the performance using Metrics

```
pd.crosstab(Y_Test,y_predict)
col_0
 Sex
   0 108 29 112
   1 33 223 35
   2 123 52 121
print(classification_report(Y_Test,y_predict))
            precision recall f1-score support
         0
                0.41 0.43 0.42
                                             249
               0.73 0.77 0.75
0.45 0.41 0.43
         1
                                             291
         2
                                             296
                                  0.54
                                             836
   accuracy
               0.53 0.54
0.54 0.54
  macro avg
                                   0.53
                                             836
               0.54
weighted avg
                        0.54
                                  0.54
                                             836
```

### Measuring the performance using metrics

```
from sklearn.metrics import mean_squared_error,mean_absolute_error
from sklearn.metrics import accuracy_score
mse = mean_squared_error(pred_test,y_test)
print("The Mean squared error is: ", mse)
rmse = np.sqrt(mse)
print("The Root mean squared error is: ", rmse)
mae = mean_absolute_error(pred_test,y_test)
print("The Mean absolute error is: ", mae)
acc = lr.score(x_test,y_test)
print("The accuracy is: ", acc)

The Mean squared error is: 3.403389401193475
The Root mean squared error is: 1.8448277429596172
The Mean absolute error is: 1.3537325298790688
The accuracy is: 0.0657871258637811
```

### 10. ADVANTAGES & DISADVANTAGES

### Advantage:

- 1. It helps students make decisions for choosing the appropriate college.
- 2. Here the chance of occurrence of error is less when compared with the existing system.
- 3. It is fast, efficient and reliable.
- 4. Avoids data redundancy and inconsistency.
- 5. Very user-friendly.
- 6. Easy accessibility of data.

### Disadvantage:

- 1. Requires an active internet connection.
- 2. System will provide inaccurate results if data entered is incorrect.

### 11. CONCLUSION

This system, which we constructed for the first time in Python employing ML algorithms and other front end languages like html, css, and java script, has proven to be more challenging than anticipated.

While it can seem straightforward to complete a few forms and process the data, there is much more that goes into choosing applications than this. Every time improvements and new features were implemented, ideas for more features or ways to make the system easier to use became obvious. A project in and of itself, balancing the completion of these required features with the suggestions for improvement as well as remembering everything that had to be done, was made possible by the fact that adding one feature made it possible to add another necessary feature. Finding out what needs to be debugged can often be a very simple task compared to debugging itself. Since so many components of the admissions system are interconnected, if an error is found on one page, it may be a display error, a problem with how the information is read from the database, or even a problem with how the information was initially stored in the database. Each time, all three must be verified. When the apparent reason for an issue is not immediately clear, this slows down the process and can be frustrating. The language used must be straightforward and simple to comprehend, and compatibility is crucial. The mobility of this system would not have been able to duplicate if it had not been created as a fully web-based application. Even if it doesn't have all the features that would have been preferred, the system performs well overall and meets initial expectations. The few mistakes that do occur are small or graphical, and the majority of the functionality that is provided works beautifully.

### 12. FUTURE SCOPE

A powerful web application can be developed where inputs are not given directly instead student parameters are taken by evaluating students through various evaluations and examining. Technical, analytical, logical, memory based, psychometry and general awareness, interests and skill based tests can be designed and parameters are collected through them so that results will be certainly accurate and the system will be more reliable to use.

### 13. APPENDIX

**Source Code** 

```
index.html
<!DOCTYPE html>
<html lang="en">
<head>
  <meta charset="UTF-8">
  <meta http-equiv="X-UA-Compatible" content="IE=edge">
  <meta name="viewport" content="width=device-width, initial-scale=1.0">
  <link rel="stylesheet" href="/styles.css">
  <title>University Admit Eligibility Predictor</title>
</head>
<body class="index-page"
    background-image: url(/home/sangamithra/Downloads/bg.jpeg);>
  <div class="index">
    <h1>University Admit Eligibility Predictor</h1>
                   <a href="/checkEligibility"><button type="button" class="btns">Check
   Eligibility</button></a>
  </div>
</body>
</html>
Chance.html
<!DOCTYPE html>
<html lang="en">
<head>
  <meta charset="UTF-8">
  <meta http-equiv="X-UA-Compatible" content="IE=edge">
  <meta name="viewport" content="width=device-width, initial-scale=1.0">
  <link rel="stylesheet" href="/css/styles.css">
  <title>University Admit Eligibility Predictor</title>
</head>
<body class="predict">
  <div class="prediction-box">
    <h1>Predicting Chance of Admission</h1>
    <div class="prediction">
      <span>Prediction:</span>
```

```
<h3 style="font-weight: 600;">You have a chance</h3>
    </div>
    <a href="/checkEligibility"><button type="button" class="btns">Back</button></a>
  </div>
</body>
</html>
Nochance.html
<!DOCTYPE html>
<html lang="en">
<head>
  <meta charset="UTF-8">
  <meta http-equiv="X-UA-Compatible" content="IE=edge">
  <meta name="viewport" content="width=device-width, initial-scale=1.0">
  <link rel="stylesheet" href="/css/styles.css">
  <title>University Admit Eligibility Predictor</title>
</head>
<body class="predict">
  <div class="prediction-box">
    <h1>Predicting Chance of Admission</h1>
    <div class="prediction">
      <span>Prediction: </span>
      <h3 style="font-weight: 600;">You don't have a chance</h3>
    </div>
    <a href="/checkEligibility"><button type="button" class="btns">Back</button></a>
  </div>
</body>
</html>
Demo2.html
<!DOCTYPE html>
<html lang="en">
<head>
  <meta charset="UTF-8">
  <meta http-equiv="X-UA-Compatible" content="IE=edge">
```

<meta name="viewport" content="width=device-width, initial-scale=1.0">

```
<link rel="stylesheet" href="/styles.css">
  <title>University Admit Eligibility Predictor</title>
</head>
<body class="home">
  <h1>University Admit Eligibility Predictor</h1>
  <form method="post" action="/predict" class="form">
    <div class="form-data">
      <label for="greScore">GRE Score:</label>
       <input name="greScore" placeholder="GRE Score" required>
    </div>
    <div class="form-data">
      <label for="toeflScore">TOEFL Score:</label>
      <input name="toeflScore" placeholder="TOEFL Score" required>
    </div>
    <div class="form-data">
      <label for="univRank">University Rank:
      <input name="univRank" placeholder="University Rank" required>
    </div>
    <div class="form-data">
      <label for="sop">SOP:</label>
      <input name="sop" placeholder="SOP" required>
    </div>
    <div class="form-data">
      <label for="lor">LOR:</label>
       <input name="lor" placeholder="LOR" required>
    </div>
    <div class="form-data">
      <label for="cgpa">CGPA:</label>
      <input name="cgpa" placeholder="CGPA" required>
    </div>
    <div class="form-data">
      <label for="research">Research:</label>
      <input name="research" placeholder="Research" required>
    </div>
    <br>
    <br>
    <button type="submit" class="btns">Submit</button>
```

```
<a href="/"><button type="button" class="btns">Back</button></a>
  </form>
</body>
</html>
```

```
Model Training
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
import seaborn as sns
data=pd.read_csv('Admission_Predict.csv')
data.head()
data.tail()
data.drop(["Serial No."],axis=1,inplace=True)
data.head()
data.info()
data.describe()
"""## Data visualization"""
plt.scatter(data['CGPA'],data['Chance of Admit '])
plt.show()
plt.bar(data['GRE Score'],data['Chance of Admit '])
plt.show()
"""## Splitting the dependent and independent column"""
x=data.iloc[:,0:7].values
x.shape
```

```
y=data.iloc[:,7:].values
y.shape
"""## Splitting the data into train and test"""
from sklearn.model selection import train test split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=10)
y train=(y train>0.5)
y train
y_test=(y_test>0.5)
y test
"""## Training and testing the model"""
X=data.drop(['Chance of Admit '],axis=1) #input data set
y=data['Chance of Admit'] #output labels
from sklearn.model selection import train test split
X train, X test, y train, y test = train test split(X, y, test size=0.15)
from sklearn.preprocessing import MinMaxScaler
scaler=MinMaxScaler()
X train[X train.columns] = scaler.fit transform(X train[X train.columns])
X \text{ test}[X \text{ test.columns}] = \text{scaler.transform}(X \text{ test}[X \text{ test.columns}])
X train.head()
from sklearn.ensemble import GradientBoostingRegressor
rgr = GradientBoostingRegressor()
rgr.fit(X train,y train)
rgr.score(X test,y test)
y predict=rgr.predict(X test)
```

```
from sklearn.metrics import mean squared error, r2 score,mean absolute error
import numpy as np
print('Mean Absolute Error:', mean absolute error(y test, y predict))
print('Mean Squared Error:', mean squared error(y test, y predict))
print('Root Mean Squared Error:', np.sqrt(mean squared error(y test, y predict)))
y_{train} = (y_{train} > 0.5)
y_test = (y_test > 0.5)
from sklearn.linear model. logistic import LogisticRegression
lore = LogisticRegression(random state=0, max iter=1000)
lr = lore.fit(X train, y train)
y pred = lr.predict(X test)
from sklearn.metrics import accuracy score, recall score, roc auc score, confusion matrix
print('Accuracy Score:', accuracy score(y test, y pred))
print('Recall Score:', recall score(y test, y pred))
print('ROC AUC Score:', roc auc score(y test, y pred))
print('Confussion Matrix:\n', confusion matrix(y test, y pred))
"""## Save the model """
import pickle
pickle.dump(lr,open('university admission.pkl','wb'))
model=pickle.load(open('university admission.pkl','rb'))
Model Evaluation
import pandas as pd
```

import pandas as pd import matplotlib.pyplot as plt import numpy as np import seaborn as sns

```
data=pd.read csv('Admission Predict.csv')
data.head()
"""## Train and Testing the model"""
x=data.drop(['Chance of Admit '],axis=1)
y=data['Chance of Admit']
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.15)
x train
y train
x_test
y_test
"""## Model Evaluation"""
from sklearn.ensemble import GradientBoostingRegressor
model = GradientBoostingRegressor()
model.fit(x_train,y_train)
model.score(x_test,y_test)
y predict=model.predict(x test)
from sklearn.metrics import mean_squared_error, r2_score,mean_absolute_error
import numpy as np
print('Mean Absolute Error:', mean absolute error(y test, y predict))
print('Mean Squared Error:', mean_squared_error(y_test, y_predict))
print('Root Mean Squared Error:', np.sqrt(mean_squared_error(y_test, y_predict)))
y train = (y train > 0.5)
```

```
y_test = (y_test>0.5)

from sklearn.linear_model._logistic import LogisticRegression
lore = LogisticRegression(random_state=0, max_iter=1000)

lr = lore.fit(x_train, y_train)
y_pred = lr.predict(x_test)

y_pred
```

### 13.2 GitHub& Project Demo Link

### GitHub link:

https://github.com/IBM-EPBL/IBM-Project-9130-1658982501

### **Project Demo Link:**

https://drive.google.com/file/d/1BWRDi-2O1ZJaz1-k-oJoUdI7TkYpX2PK/view