UNIVERSITY ADMIT ELIGIBILITY PREDICTOR

PROJECT REPORT

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1. INTRODUCTION

The world's business sector is escalating and is constantly seeking information and experiences that are commonly beneficial to individuals. Young specialists who need to stay in their current positions are always looking for advanced degrees to help them address their skills and information. As such, the number of her sophomores applying for graduation exams has increased over the past decade. One of her main concerns is getting into fantasy her university. You can see that undergraduates are actually choosing to get their education at prestigious universities. Furthermore, when it comes to international alumni, the United States is the main trend for most of them. The most prestigious universities offer a wide range of courses accessible in any order, exceptionally accredited teaching and education programs, an international second Research scholarships for degrees are available.

According to Gauges, more than 4,444 of her 10 million international sophomores are enrolled in her 4,200+ colleges and universities, both private and public. In general, the number of undergraduates concentrated in America comes from Asian countries such as India, Pakistan, Sri Lanka, Japan and China. Select the United Kingdom, Germany, Italy, Australia, Canada as well as the United States. These countries are witnessing a rapid increase in the number of individuals seeking more advanced investigations. The basic reason why sophomores go on to master's programs in foreign graduate schools is that the number of vacancies is low and the number of people in these positions in each country is huge. This has led many professional undergraduates to pursue postgraduate studies. You can see that there are quite a few bachelor's degrees and master's degrees in computer science at US universities. The focus of this study applies to these undergraduate degrees. Many schools in the US follow comparative requirements for

undergraduate accreditation. Schools consider several variables, including placement in fitness assessments and school performance ratings. English rankings are determined by exposure in English proficiency tests such as TOEFL and IELTS.

The University's Admissions Advisory Board makes decisions regarding the acceptance or rejection of specific young researchers based on the general profile of the applicant's application. Records recorded with this company are marked with informative areas. Acknowledgment is a 400-row data set containing seven different autonomic factors. ie

- Graduate Record Examination 1 (GRE) score. The score consists of 340 foci.
- English as a Foreign Language (TOEFL) test score. It consists of 120 priority areas.
- Uni.Rating. Shows the position of colleges offering bachelor's degrees among various colleges. Your score will be out of 5.
- Statement of Purpose (SOP), a record written to reveal the life, motivations and inspirations of a selected degree/college applicant. The score consists of five focal points.
- The strength of a letter of recommendation (LOR) verifies the applicant's professional experience, falsifies validity, supports certainty, and guarantees your competence. The score consists of five focal points.
- Undergraduate GPA (CGPA) from 10.
- Research experience (either 0 or 1) that could support the application, such as distributing research papers at conferences or filling out as a right-hand exam for university faculty. One ward variable can be anticipated which is possibility of affirmation, that is as per the input given will be going from 0 to 1.

1.1 PROJECT OVERVIEW

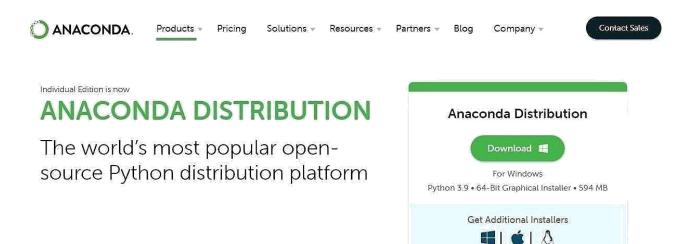
PRE-REQUISTIES

Anaconda Installation:

Anaconda is a distribution of the Python and R programming languages for scientific computing that aims to simplify package management and deployment. The distribution includes data science packages suitable for Windows, Linux, and macOS. Developed and maintained by Anaconda. Founded in 2012 by Peter Wang and Travis Olyphant. As Anaconda, also known as Anaconda Distribution or Anaconda Individual Edition, the company's other products include his Anaconda Team Edition and Anaconda Enterprise Edition, neither of which are free.

WAY TO INSTALL ANACONDA:

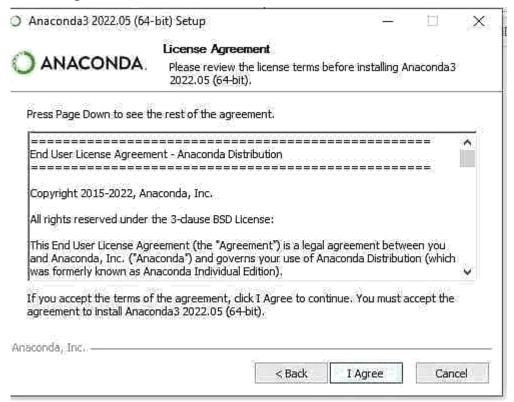
STEP 1: Download and Anaconda



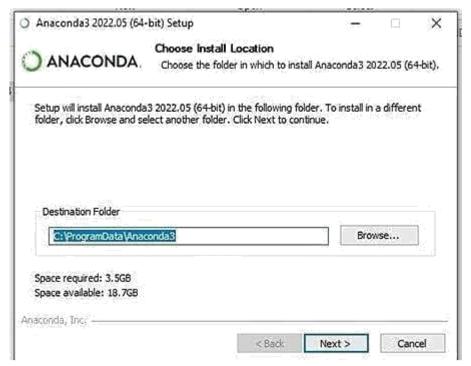
STEP 2: Install the Anaconda



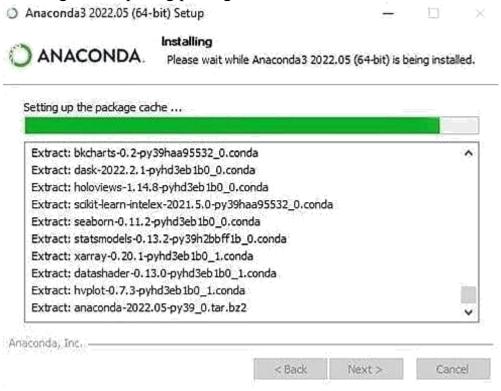
STEP 3: Click I Agree



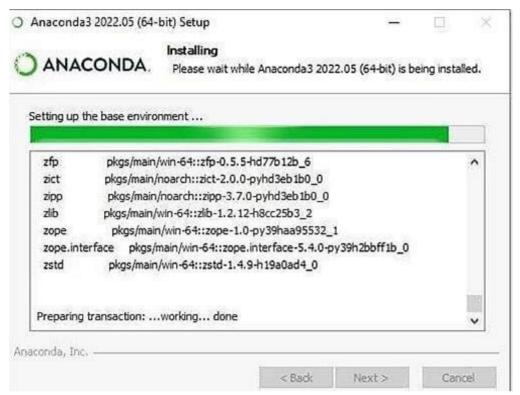
STEP 4: Choose the Installation Location



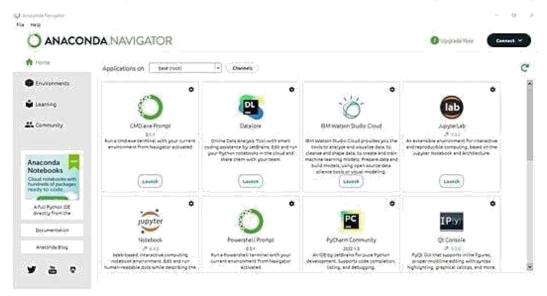
STEP 5: Installing the Requiring packages



STEP 6: Setting up the base environment

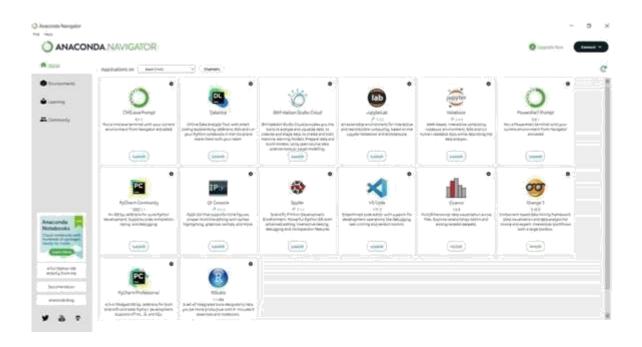


STEP 7: Successfully Installed and check the Anaconda Navigator working or not

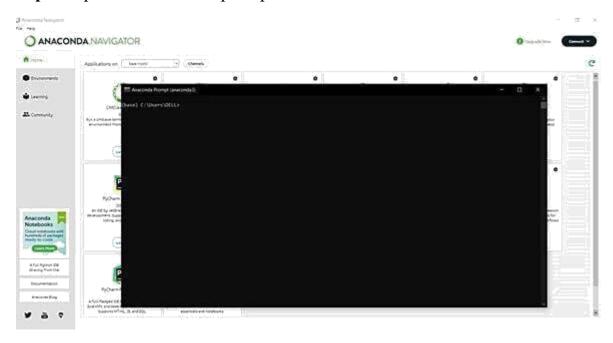


Python packages installation:

Step 1: Open the anaconda navigator in the start menu



Step 2: Open the CMD.exe prompt



Step 3: Install the NUMPY package

To enter the numpy package enter the command in the CMD.exe

Command: Pip install numpy

Numpy:

This package is used to perform numerical computations. This package comes preinstalled with Anaconda. NumPy is used for manipulating arrays. NumPy stands for Numerical Python.



Step 4: Install the pandas package.

To enter the pandas package enter the command in the CMD.exe

Command: Pip install pandas

Pandas:

Pandas is one of the most widely used Python libraries for data science. It provides powerful and easy-to-use structure and data analysis tools. This package comes pre-installed with Anaconda. An open source library built on top of the NumPy library. A Python package that provides various data structures and operations for working with numerical data and time series. Mainly, it's common for data to be imported and analyzed much easier. Pandas is fast, providing users with high performance and productivity.

```
Anaconda Prompt (anaconda3) - pip install pandas

(base) C:\Users\DELL>pip install pandas

Requirement already satisfied: pandas in c:\users\dell\anaconda3\lib\site-packages (1.4.2)

Requirement already satisfied: python-dateutil>=2.8.1 in c:\users\dell\anaconda3\lib\site-packages (from pandas) (2.8.2)

Requirement already satisfied: numpy>=1.18.5 in c:\users\dell\anaconda3\lib\site-packages (from pandas) (1.21.5)

Requirement already satisfied: pytz>=2020.1 in c:\users\dell\anaconda3\lib\site-packages (from pandas) (2021.3)

Requirement already satisfied: six>=1.5 in c:\users\dell\anaconda3\lib\site-packages (from python-dateutil>=2.8.1->panda
s) (1.16.0)
```

Step 5: Install the Matplotlib package.

To enter the Matplotlib package enter the command in the CMD.exe

Command: Pip install Matplotlib

Matplotlib:

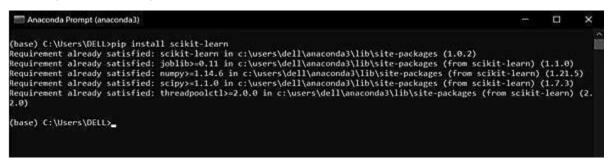
Matplotlib is a comprehensive library for creating static, animated and interactive visualizations in Python. This package comes pre-installed with Anaconda. Matplotlib is a nice visualization library in Python for 2D plotting of arrays. Matplotlib is a cross-platform data visualization library based on NumPy arrays and designed to work with the wider SciPy stack. Introduced by John Hunter in 2002.



Step 6: Install the Scikit-learn package.

To enter the Scikit- learn package enter the command in the CMD.exe

This is a machine learning library for the Python programming language. This package comes pre-installed with Anaconda. Scikit Learn in Python is primarily used to focus on modeling in Python. It was only focused on modeling, not loading data.



Step 7: Install the Flask package.

To enter the Flask package enter the command in the CMD.exe

Command: Pip install Flask

Flask:

Flask is a lightweight WSGI web application framework Flask is a web application framework written in Python. It is developed by Armin Ronacher, who leads an international group of Python enthusiasts called Pocco. Flask is based on the WSGI toolkit tools and the Jinja2 template engine. Both are Pocco projects.

```
(base) C:\Users\DELL>pip install flask
Requirement already satisfied: flask in c:\users\dell\anaconda3\lib\site-packages (1.1.2)
Requirement already satisfied: click>5.1 in c:\users\dell\anaconda3\lib\site-packages (from flask) (8.0.4)
Requirement already satisfied: Werkzeug>=0.15 in c:\users\dell\anaconda3\lib\site-packages (from flask) (2.0.3)
Requirement already satisfied: Jinja2>=2.10.1 in c:\users\dell\anaconda3\lib\site-packages (from flask) (2.11.3)
Requirement already satisfied: itsdangerous>=0.24 in c:\users\dell\anaconda3\lib\site-packages (from flask) (2.0.1)
Requirement already satisfied: colorama in c:\users\dell\anaconda3\lib\site-packages (from click>=5.1->flask) (0.4.4)
Requirement already satisfied: MarkupSafe>=0.23 in c:\users\dell\anaconda3\lib\site-packages (from linja2>=2.10.1->flask
) (2.0.1)
(base) C:\Users\DELL>
```

PROJECT FLOW

You will go through all the steps mentioned below to complete the project.

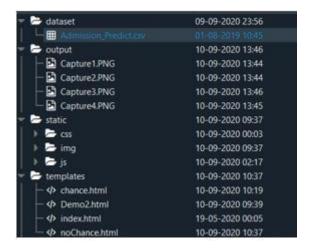
- User interacts with the UI (User Interface) to enter Data
- The entered data is analyzed by the model which is integrated
- Once model analyses the input the prediction is showcased on the UI To accomplish this, we have to complete all the activities and tasks listed below
 - Data Collection.
 - Collect the dataset or Create the dataset
 - Data Preprocessing.
 - Import the Libraries.
 - Importing the dataset.
 - Checking for Null Values.
 - Data Visualization.
 - Taking care of Missing Data.
 - Label encoding.
 - One Hot Encoding.
 - Feature Scaling.
 - o Splitting Data into Train and Test.
 - Model Building
 - Training and testing the model
 - Evaluation of Model
 - Application Building
 - Create an HTML file
 - o Build a Python Code

PROJECT OBJECTIVES

- To understand regression and classification problems.
- To grab insights from data through visualization.
- Applying different Machine Learning algorithms to determine the probability of acceptance in a particular university.
- Evaluation metrics build a web application using the Flask framework

PROJECT STRUCTURE

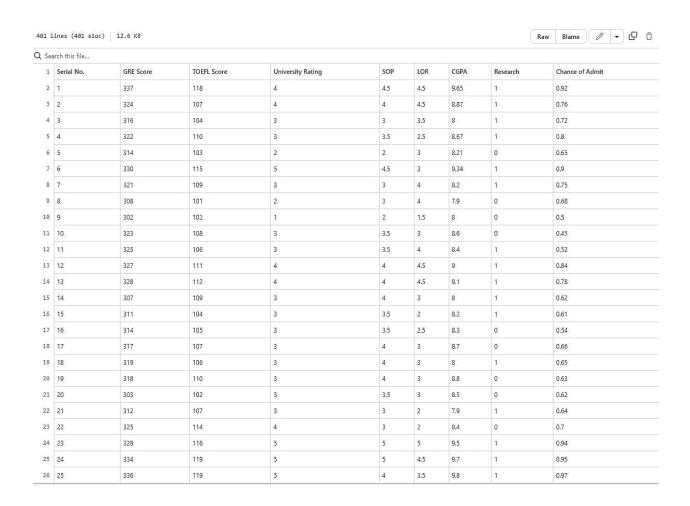
Create a Project folder that contains files as shown below



DATA COLLECTION

The path to common information varies by project type. ML projects use real-time information. Information indexes can be collected from a variety of sources such as documents, data sets, sensors, and other sources, using free information collection from the Internet. Kaggle and the UCI Machinelearning Repository are the most commonly used repositories for sorting.

Collect the dataset or Create the dataset.



DATA PRE-PROCESSING

Importing the Libraries:

• It is important to import all the necessary libraries such as pandas, numpy, matplotlib.

- **Numpy:** It is an open-source numerical Python library. It contains a multi-dimensional array and matrix data structures. It can be used to perform mathematical operations on arrays such as trigonometric, statistical, and algebraic routines.
- **Pandas:** It is a fast, powerful, flexible and easy to use open-source data analysis and manipulation tool, built on top of the Python programming language.
- **Seaborn:** Seaborn is a Python data visualization library based on matplotlib. It provides a high-level interface for drawing attractive and informative statistical graphics.
- **Matplotlib:** Visualisation with python. It is a comprehensive library for creating static, animated, and interactive visualizations in Python

Reading the Datasets:

1. You might have your data in .csv files, .excel files

Let's load a .csv data file into pandas usingread_csv() function. We will need to locate the directory of the CSV file at first

```
#read_csv is a pandas function to read csv files
data = pd.read_csv('Admission_Predict.csv')
```

If your dataset is in some other location, Then see below command Data=

pd.read_csv(r"File_location/filename.csv")

Note: r stands for "raw" and will cause backslashes in the string to be interpreted as actual backslashes rather than special characters.

Our Dataset Admission_Predict contains following Columns 1.Serial No.

- 2. GRE Score
- 3. TOEFL Score
- 4. University Rating
- 5. SOP
- 6. LOR
- 7. CGPA
- 8. Chance of Admit

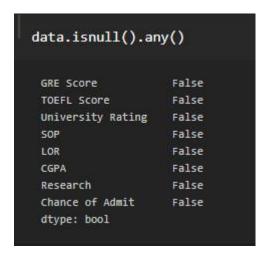
Handling Missing Values:

After loading it is important to check the complete information of data as it can indicate many of the hidden information such as null values in a column or a row

Check for the null values. if it is present then the following steps can be performed

- Imputing data using the Imputation method in sklearn.
- Filling NaNvalues with mean, median, and mode using

fillna() method. You can check the null values with the function isnull().any()



If the dataset contains null values then the above functions return as true. But if you look at the dataset you can observe that the dataset does not have any null values.

You can also check the number of null values present in the columns by the using isnull().sum() function. As we don't have categorical data then we can skip the steps of label encoding and one-hot encoding

Data Visualization:

Data visualization is where a given dataset is presented in a graphical format. It helps the detection of patterns, trends and correlations that might go undetected in text-based data. Understanding your data and the relationship present within it is just as important as any algorithm used to train your machine learning model. Machine learning models will perform poorly on data that wasn't visualized and understood properly.

To visualize the dataset we need libraries called Matplotlib and Seaborn. The Matplotlib library is a Python 2D plotting library that allows you to generate plots, scatter plots, histograms, bar charts etc.

Splitting Dependent And Independent Columns:

We need to split our dataset into the matrix of independent variables and the vector or dependent variable. Mathematically, Vector is defined as a matrix that has just one column.

• To read the columns, we will useiloc of pandas (used to fix the indexes for selection) which takes two parameters — [row selection, column selection].

Let's split our dataset into independent and dependent variables.

```
x=data.iloc[:,0:7].values
x

y=data.iloc[:,7:].values
y
```

From the above code ":" indicates that you are considering all the rows in the dataset and "0:7" indicates that you are considering columns 0 to 7 such as year, month, and day as input values and assigning them to variable x. In the same way in the second line ":" indicates you are considering all the rows and "7:" indicates that you are considering only the last column as output value and assigning them to variable y.

Let's Check the shape of x and Y

```
x.shape
(1991, 7)
y.shape
(1991, 1)
```

• You can see in x we have 1991 rows with 7 columns and y has 1 column with the same number of rows

Splitting The Data Into Train And Test:

To train the model, first split the model into two segments: "training data" and "testing data". The classifier is trained using a 'training data set' and the performance of the classifier is tested on a non-fitting 'test data set'.

Training Set: The training Set is material for computers to learn how to process data. The AI uses computation to do the training part. The training dataset is used to learn and tune the classifier parameters.

Test set: A set of unseen data used solely to evaluate the performance of thefully displayed classifier.

When you are working on a model and you want to train it, you obviously have a dataset. But after training, we have to test the model on some test dataset. For this, you will need a dataset that is different from the training set you used earlier. But it might not always be possible to have so much data during the development phase. In such cases, the solution is to split the dataset into two sets, one for training and the other for testing.

The train-test split is a technique for evaluating the performance of a machine learning algorithm.

- Train Dataset:Used to fit the machine learning model.
- Test Dataset:Used to evaluate the fit machine learning model.

```
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)
```

In general, you can allocate 80% of the dataset to the training set and the remaining 20% to the test set. We will create 4 sets

- 1. x train
- 2. x test
- 3. y_train
- 4. y_test

There are a few other parameters that we need to understand before we use the class:

- test_size: this parameter decides the size of the data that has to be split as the test dataset. This is given as a fraction. For example, if you pass 0.5 as the value, the dataset will be split 50% as the test dataset and remaining a train dataset
- random_state: here you pass an integer, which will act as the seed for the random number generator during the split. Or, you can also pass an instance of the Random_state class, which will become the number generator. If you don't pass anything, the Random_state instance used by np.random will be used instead.

1.2 PURPOSE

This is the project for a new web-based University Admit Eligibility Predictor. Predictor is an ML based application that asks for the users to input their academic transcripts data and calculates their chances of admission into the University Tier that they selected. It also provides an analysis of the data and shows how chances of admissions can depend on various factors. This document describes the scope, objectives and goals of the system. In addition to describing the non-functional requirements, this document models the functional requirements with use cases, interaction diagrams and class models. This document is intended to direct the design and implementation of the target system in an object-oriented language.

2. LITERATURE SURVEY 2.1

Existing Problem

It's almost admission season and I've couple of friends who are in panic mode waiting for a call from the universities they've applied at.

This made me think — How can we predict whether a student will get an admit or not? What are the parameters for selection? Can it be mathematically expressed?

All of these questions started popping up. This is the main existing problem.

2.2 References

A UNIVERSITY ADMISSION PREDICTION SYSTEM USING STACKED ENSEMBLE LEARNING

AUTHOR: S. Sridhar, S. Mootha and S. Kolagati OVERVIEW: "A University Admission Prediction System using Stacked Ensemble Learning," 2020 Advanced Computing and Communication Technologies for High Performance Applications (ACCTHPA), 2020, pp. 162-167, doi: 10.1109/ACCTHPA49271.2020.9213205. Abstract: For an aspiring graduate student, shortlisting the universities to apply is a difficult problem. Since an application is extremely dynamic, students often tend to wonder if their profile matches the requirement of a certain university. Moreover, the cost of applying to a university is extremely high making it critical that students shortlist universities based on their profile. A university admission prediction system is quite useful for students to determine their chances of acceptance to a specific university. The system could make use of data related to previous applicants to various universities and their admit or reject status. Earlier models of such prediction systems suffer from several drawbacks such as not considering important parameters like GRE (Graduate Record Exam) scores or research experience. Further, the accuracy reported by earlier models is also not sufficiently high. In this paper, a stacked ensemble model that predicts the chances of admit of a student to a particular university has been proposed. The proposed model takes into consideration various factors related to the student including their research experience, industry experience etc. Further, the system proposed has been evaluated against various other machine learning algorithms including other deep learning methods. It is observed that the proposed model easily outperforms all other models and provides a very high accuracy. Observation: An effective method has been proposed to predict the chances of a student being admitted to a specific university. In addition, we have compared the performance of various machine learning algorithms to

the proposed method in predicting admits.

ACCURACY: 91%

URL: https://ieeexplore.ieee.org/document/9213205

TITLE: PREDICTION PROBABILITY OF GETTING AN ADMISSION INTO A UNIVERSITY USING ML

AUTHOR: A. Sivasangari, V. Shivani, Y. Bindhu, D. Deepa, R. Vignesh OVERVIEW: In the present conditions, students regularly have difficulty finding a fitting institution to pursue higher studies based on their profile. There are some advisory administrations and online apps that recommend universities but they ask huge consultancy fees and online apps are not accurate. So, the aim of this research is to develop a model that predict the percentage of chances into the university accurately. This model provides also the analysis of scores versus chance of prediction based on historical data so that students can understand whether their profile is suitable or not. The proposed model uses linear regression and random forest algorithms but cat boost algorithm is giving highest accuracy. ACCURACY: 95%

URL: https://ieeexplore.ieee.org/document/9418279

TITLE: PREDICTING THE POST GRADUATE ADMISSIONS USING CLASSIFICATION TECHNIQUES

AUTHOR: Selvaprabu Jeganathan, Saravanan Parthasarathy and P. M. Ashok Kumar OVERVIEW: Decision making by applying data mining methods is being used in many service organizations. Educational bodies gradually started to use the business intelligence techniques to identify the current progress in their institutions. Numerous factors which have an impact in academia will be vivid to the educationalists while applying data mining techniques on the academic data. By employing the data mining methodologies, we could identify different patterns which aid institutions to take strategic decisions to improve the students' academic performance. Potential graduate students will have a dilemma on identifying the universities for their post graduate admissions and on the other hand an average graduate student would be uncertain on getting post graduate admission in a reputed university based on their academic scores. In this study, we applied the classification techniques such as Logistic Regression, KNN Classification, Support Vector Classification, Naive Bayes Classification, Decision Tree Classification and Random Forest Classification on the given academic admission dataset.

ACCURACY: 99%

URL: https://ieeexplore.ieee.org/document/939681

2.3 Problem Statement Definition

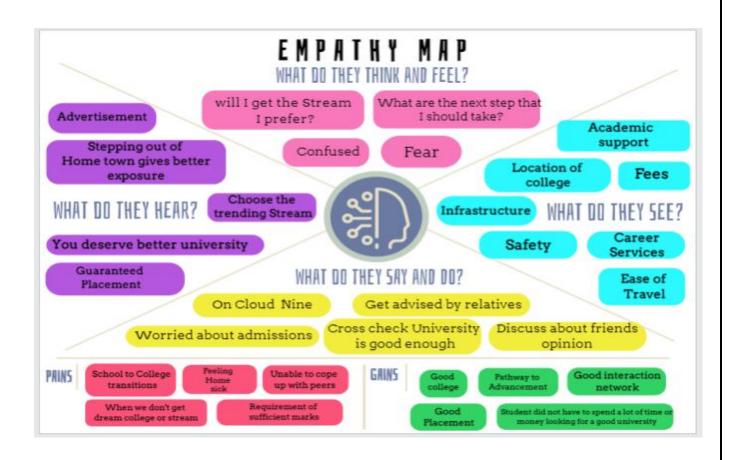
Problem Statement	I am (Customer)	I'm trying to	But	Because	Which makes me feel
PS-1	A student	Enroll for a PG course in the University	I am unaware of eligible cut off	The calculation for the cut off is not available in the website	university to
PS-2	Administator	To download all the data from the students	Details furnished by the students are incomplete	There is the problem in the seerver because of huge volume if data	Frustated and not able to download the documents

3. IDEATION AND PROPOSED SOLUTION

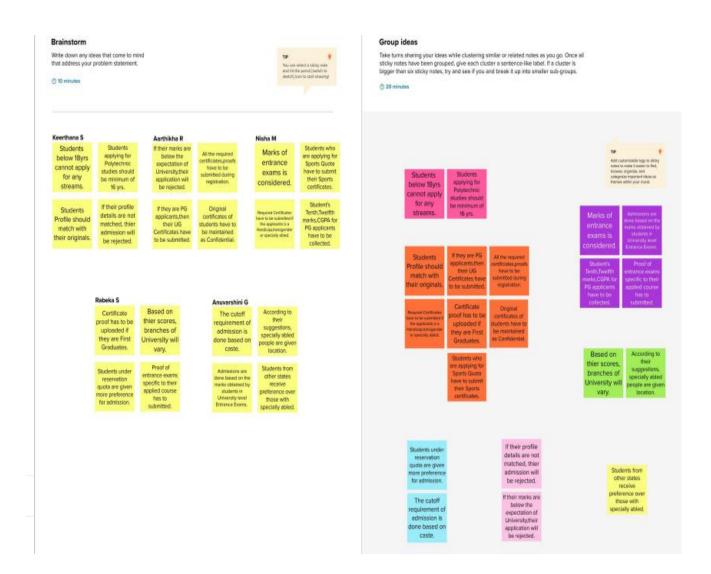
Ideation is the process where you generate ideas and solutions through sessions such as Sketching, Prototyping, Brainstorming, Brainwriting, Worst Possible Idea, and a wealth of other ideation techniques. Ideation is also the third stage in the Design Thinking process. In this project the ideation phase consist of,

- Empathy Map
- Brainstorming
- Proposed Solution
- Problem Solution Fit

3.1Empathy Map



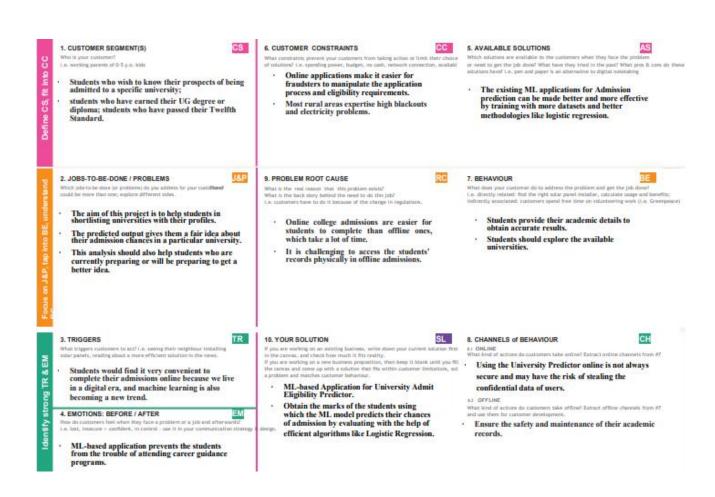
3.2 Ideation & Brainstorming



3.3 Proposed Solution

S.No	Parameter	Description
1.	Problem Statement (Problem to be solved)	Generally as the students don't have much idea about the procedures, requirements and details of the universities in the India they seek help from the education consultancy firms to help them successfully secure admission in the universities which are best suitable for their profile, for this they have to invest huge amount of money as consultancy fees.
2.	Idea / Solution description	Providing an as accurate as possible prediction for the student's chances of admissions to the universities of their choice based on their academic transcripts with the help of an Web Application and Machine Learning algorithms
3.	Novelty / Uniqueness	It seems that there are no application or web application which is common for all over India for predicting the eligibility of an student for Indian university, so the proposed application will contain the eligibility criteria of all the Indian universities.
4.	Social Impact / Customer Satisfaction	It helps student for making decision for choosing a right college. Allows students to check ever college eligibility criteria. Avoid middle man and make direct connection between Students and Universities.
5.	Business Model (Revenue Model)	Like most areas of the educational world these days, technology is forcing higher education institutions to do more with less. Institutions are under increasing pressure to admit more students, retain these students, and do their best to ensure student success. Facing this pressure, tech-savvy institutions can benefit greatly from predictive analytics and predictive models to help achieve their goals.
6.	Scalability of the Solution	This will also help you to finalize your dream schools with a realistic road map, with the help of information coupled with a bit of reality check on your academic scores, work experience, your eminence over your peers. we have connoisseurs who shall work with you to amplify your prospects ofreceiving offers by ensuring that universities you apply.

3.4 Problem Solution Fit



4. REQUIREMENT ANALYSIS

4.1 Functional Requirements

Following are the functional requirements of the proposed solution.

FR No.	Functional Requirement	Sub Requirement (Story / Sub-Task)		
FR-1	User Registration	Registration through Form Registration through Gmail		
FR-2	User Confirmation	Confirmation via Email Confirmation via OTP		
FR-3	User Data Collection	He following details of Students' Score are collected: 1.HSC 2.SSLC 3.CGPA if their PG Applicants.		
FR-4	Evaluation	Using ML algorithms to analyse the data entered by the students and testing the developed ML model with the supplied data.		
FR-5	Prediction	Prediction is done based on the result of evaluation, the List of Universities for which the students are eligible to apply will be displayed.		
FR-6	Output	Based on their eligibility, students move forward with the admissions procedure to the predicted university and course.		

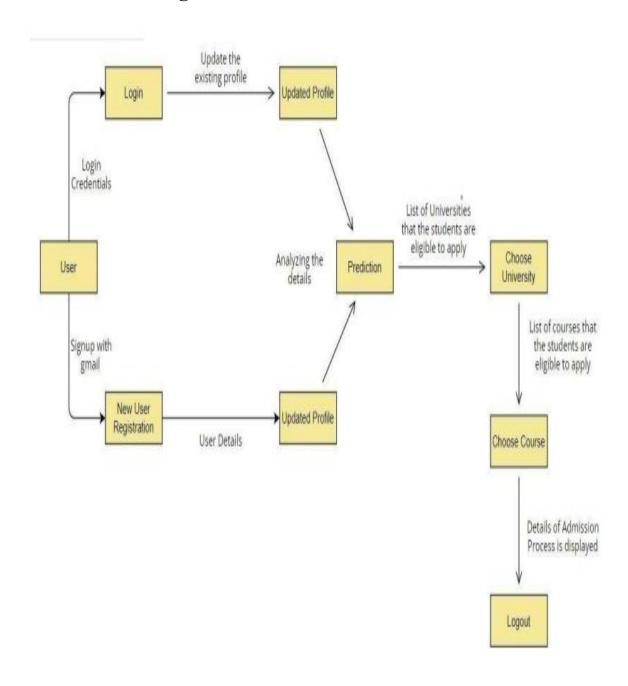
4.2 Non-functional Requirements:

Following are the non-functional requirements of the proposed solution.

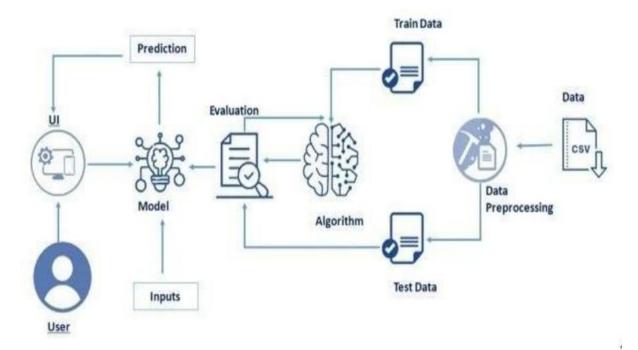
NFR No.	Non-Functional Requirements	Description
NFR-1	Usability	Interactive and Effective UI Visualization of Progress Customer Satisfaction Ease of Learning
NFR-2	Security	Frequent Updates using the Customers' feedback. Automatic Logout when the app is not in use to prevent unauthorized access to the user's account.
NFR-3	Reliability	The predictor system will be consistent in order for the system to produce trustworthy and accurate outcomes.
NFR-4	Performance	As logistic regression is applied to develop, performance will be more effective.
NFR-5	Availability	Users will be able to access the system predictor at any time, anyplace, as needed.
NFR-6	Scalability	It can handle any amount of data and perform many computations in a cost - effective and time - saving way.

5. PROJECT DESIGN

5.1 Data Flow Diagram



5.2 Solution & Technical Architecture

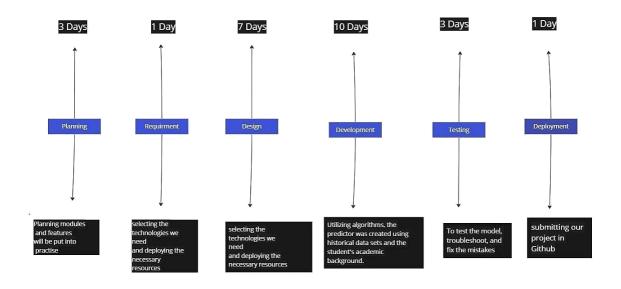


5.3 User Stories

Error debids, Christies University, Depost Generation, Psycliding Generation of Agriculture Agriculture	Entice How does warmone industry faccore owners of the process?	Enter What do poople apprinting on they begin the process?	Engage In the case promotes shall response, whell response, whell	Exit What do poolph Typicarly experience as the process finithes? Extend What happen
tik Steps What about the position for group! hypersity repersised?	Committee Spen, Series	Against Law Specialists Committeesing The committees of the commi	Sections: State of Course	Imperior Freehoods Althous controlling for processing and process
Interactions What stranscious is they have at each beg doing the way? * People Woo do they see at lab to? * Please Where are deep? * Things What digatal standpoints or ginyated objects would they use?	The manufacture of the control of th	The second state of the se	The property of the property o	Salak ke saya salakan Kalak ke saya ke salak sal
Goals & motivations Ac each strp, where is possion's private good or meta-scale? [**Neg me.," or "Help me eved"]	maj un equipus de la librar en en estado en el la librar de la librar en el la librar de la librar en el la li	Special color deposit provide color provide color deposit	Heaving for the control of the contr	According to the second
Positive moments What inspiredoes nighted person feel exception, teach recovering, designation or excellent	Josephanes (Promotores and Control of Contro	Later interests 1. Annual and	The act in the control of the contro	Common Vigoria Annual Common Vigoria Common
Negative moments When visco does elected receive statisfications, containing, expering, sootly, or time consuming?	To Committee Com	one factories program, section for and section for and section for any other program of the prog	Assessing the second of the se	We manage dis- paration of a source in the federal source in the federal source in the source in the source in the source in the source in the source in the source in the source in the source in the
Areas of opportunity How might we make each step better? What does do se fee? What hose others supported?	Pening for forces to design of sold of the pening of the pen	so dang di sabiliti da dang sari vidan (na sari sabiliti da dang sari vidan (na sari sabiliti da dang sari sab	biomissionics broaders we would be seen to consider the second of the se	To any consist of sections of

6. PROJECT PLANNING & SCHEDULING

6.1 Sprint Planning & Estimation



mirc

6.2 Sprint Delivery Schedule

Sprint	Functiona l Requirem ent	User Story No.	User Story / Task	Story Point s	Priori ty	Team Members
Sprint-	Registration	USN-1	As a user, I can register for the application by entering my email, password, and confirming my password	2	High	DEVADHARSHINI M
Sprint-		USN-2	As a user, I will receive confirmation email once I have registered for the application	1	High	DHARANI V
Sprint-2		USN-3	As a user, I can register for the application	2	Low	DIVYA J

			through Facebook			
Sprint-		USN-4	As a user, I can register	2	Mediu	DIVYA
1			for the application		m	BHARATHI S
			through Gmail			
Sprint-	login	USN-5	As a user, I can log into	1	High	DEVA DHARSHINI
1			the application by			M
			entering email &			
			password			
	dashboard					

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

7. CODING & SOLUTIONING

7.1 Feature

The new feature will predict the chances in the admission of the university. The feature was designed in the html code connected with app.py as the backend.

```
<!DOCTYPE html>
0hdf<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="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:</label>
      <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>
```

7.3 Database Schema

The database used here in this project was Admission_Predict.csv. The sample screenshot of the database are,

ø	NIAME.								
s	initial Miss.	GRC Scene	100X State	University Rating	509	SON.	CGPA	Equatric .	Chinose of Adriet
9		987	110	4	45	4.5	9,65	1.	0.82
2		384	887	4	4	8.5	8.87	1	0.76
3	i	316	104	3	3	0.5	4.	1	0.72
4		322	110	1	88	23	487	т	0.8
ý	r -	394 -	165	2	2.	3	8.25	σ-	0.65
		380	110	3	49	18	9.84	1	0.8:
7		301	100	3	á	E	8.2	7.	0.2%
8		See	101	3	4	-14	7.0	9	0.89
		362	162	1	25	13	18		0,5
19	0.	303.	180	Ti .	3.5	3	8.6	8	0.45
3	R.	385	100-	3	3,5	[#	4.4	1	0,50:
15	t t	327	111	4	4	(48	-	1	0.86
1	8	508	312:	á.	a	45	9.1	Ť	0.76
ń	é-	347	199	4	rit.	10	à	1	0.60
12	4	Taen	196	3	3.5	8	43	4	9.81
10	6.	294	YES	1	3.5	2.5	83		0.54
1	T.	207	187	3	4	3	4.7		0.86
ú	2-	\$19	150	3	4	38	a.	1	0.65
-	9	518	110	3:	-4	5	4.0		0.63
25	5	563	ing	9	38	14	4.9	8	0.62
2	H.	912	107	3	.8	2	7.9	7	0.68
2	0:	2015	fisa:	á.	3:	2	44	2	0.7
7	Q)	200	116.	5:	5	5	4.5	1	0.84
. 3	94.	304	1100	3	4	43	9.7	7	0.89
2	5-	336-	119	5	4	555	9.6	1	OST

8. TESTING

User Acceptance Testing

• 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).

• Defect Analysis

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

Solution	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 Reproduce	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

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

9. RESULTS

9.1 Performance Metrics

d.cr	ssta	b(Y_	Test,	y_pred	ict)		
col_0	0	1	2				
Sex							
0	108	29	112				
1	33	223	35				
2	123	52	121				
orint(clas	sifi				,y_predict	
orint(clas	sifi				,y_predict f1-score)) support
orint(clas	sifi 0	preci	sion		f1-score	
orint(clas	0	preci	sion 0.41	recall	f1-score 0.42	support
orint(clas		preci	sion 0.41	recall 0.43	f1-score 0.42	support 249
	clas	0 1 2	preci	0.41 0.73	recall 0.43 0.77	f1-score 0.42 0.75	support 249 291
ac		0 1 2	preci	0.41 0.73	recall 0.43 0.77	f1-score 0.42 0.75 0.43	support 249 291 296

10. ADVANTAGES&DISADVANTAGES 10.1

Advantages

- It helps student for making decision for choosing a right college.
- Here the chance of occurrence of error is less when compared with the existing system.
- It is fast, efficient and reliable.
- Avoids data redundancy and inconsistency.
- Very user-friendly.
- Easy accessibility of data.

10.2 Dis-Advantages

- Required active internet connection.
- System will provide inaccurate results if data entered incorrectly.

11. CONCLUSION

This system ,being the first we have created in Python using ML algorithms and other front end languages such as html, css, java script, has proven more difficult than originally imagined. While it may sound simple to fill out a few forms and process the information, much more is involved in the selection of applicants than this. Every time progress was made and features were added, ideas for additional features or methods to improve the usability of the system made themselves apparent. Furthermore, adding one feature meant that another required feature was now possible, and balancing completing these required features with the ideas for improvement as well as remembering everything that had to be done was a project in itself. Debugging can sometimes be a relatively straight forward process, or rather rather finding out what you must debug can be. Since so many parts of the admissions system are integrated into one another, if an error occurs on one page, it may be a display error, for example; it may be the information is not correctly read from the database; or even that the information is not correctly stored in the database initially, and all three must be checked on each occasion. This slows down the process and can be frustrating if the apparent cause of a problem is not obvious at first. Language used must be simple and easy to understand and compatibility is paramount. If this system were not designed as an entirely web based application, it would not have been possible to recreate its current state of portability. Overall, the system performs well, and while it does not include all of the features that may have been desired, it lives up to initial expectations. The majority of features that are included work flawlessly and the errors that do exist are minor or graphical.

12. FUTURE SCOPE

The future scope of this project is very broad. Few of them are:

- This can be implemented in less time for proper admission process.
- This can be accessed anytime anywhere, since it is a web application provided only an internet connection.
- The user had not need to travel a long distance for the admission and his/her time is also saved as a result of this automated system.

13. APPENDIX

13.1 Source Code

PYTHON CODE

Uploading the python code,

Import Libraries

In [1]:

import numpy as np

import pandas as pd

import matplotlib.pyplot as plt

import seaborn as sns

Data Collection

In [16]:

import os, types

import pandas as pd

from botocore.client import Configimport ibm boto3

def iter (self): **return** 0

@hidden_cell# The following code accesses a file in your IBM Cloud Object Storage. It includes your credentials.# You might want to remove those credentials before you share the notebook.cos_client = ibm_boto3.client(service_name='s3',

 $ibm_api_key_id = IYMIjb2yXoDD9i4rh2MgnyvRiI_9CoyB3MeIAbEk4JMK',$

ibm_auth_endpoint="https://iam.cloud.ibm.com/oidc/token",

config=Config(signature_version='oauth'),

endpoint_url='https://s3.private.us.cloud-object-storage.appdomain.cloud')

bucket = 'ibmuniv-donotdelete-pr-xvu5spl8e33fmm'object key = 'Admission Predict.csv'

body = cos_client.get_object(Bucket=bucket,Key=object_key)['Body']# add missing __iter__ method, so pandas accepts body as file-like objectif not hasattr(body, "__iter__"):

body.__iter__ = types.MethodType(__iter__, body)

data = pd.read_csv(body)data.head()

Out[16]:

	Serial No.	GRE Score	TOEFL Score	University Rating	SOP	LOR	CGPA	Research	Chance of Admit
0	1	337	118	4	4.5	4.5	9.65	1	0.92
1	2	324	107	4	4.0	4.5	8.87	1	0.76
2	3	316	104	3	3.0	3.5	8.00	1	0.72
3	4	322	110	3	3.5	2.5	8.67	1	0.80
4	5	314	103	2	2.0	3.0	8.21	0	0.65

In [41]:

data.describe()

Out[41]:

	Serial No.	GRE Score	TOEFL Score	Universit y Rating	SOP	LOR	CGPA	Research	Chance of Admit
coun t	400.00000	400.00000 0	400.00000 0	400.00000 0	400.00000 0	400.00000 0	400.00000 0	400.00000 0	400.00000 0
mea n	200.50000	316.80750 0	107.41000 0	3.087500	3.400000	3.452500	8.598925	0.547500	0.724350
std	115.61430 1	11.473646	6.069514	1.143728	1.006869	0.898478	0.596317	0.498362	0.142609
min	1.000000	290.00000 0	92.000000	1.000000	1.000000	1.000000	6.800000	0.000000	0.340000
25%	100.75000 0	308.00000	103.00000 0	2.000000	2.500000	3.000000	8.170000	0.000000	0.640000
50%	200.50000	317.00000 0	107.00000 0	3.000000	3.500000	3.500000	8.610000	1.000000	0.730000
75%	300.25000 0	325.00000 0	112.00000 0	4.000000	4.000000	4.000000	9.062500	1.000000	0.830000
max	400.00000	340.00000 0	120.00000 0	5.000000	5.000000	5.000000	9.920000	1.000000	0.970000

In [17]:

data.info()

RangeIndex: 400 entries, 0 to 399

Data columns (total 9 columns):

Column Non-Null Count Dtype

--- ----- -----

0 Serial No. 400 non-null int64

1 GRE Score 400 non-null int64

2 TOEFL Score 400 non-null int64

3 University Rating 400 non-null int64

4 SOP 400 non-null float64

5 LOR 400 non-null float64

6 CGPA 400 non-null float64

7 Research 400 non-null int64

8 Chance of Admit 400 non-null float64

dtypes: float64(4), int64(5) memory usage: 28.2 KB

Handling Missing Values

In [18]:

data.isnull().any()

Out[18]:

Serial No. False

GRE Score False

TOEFL Score False

University Rating False

SOP False

LOR False

CGPA False

Research False

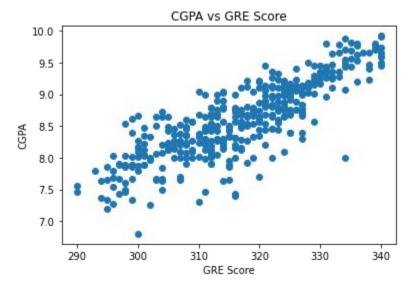
Chance of Admit False

dtype: bool

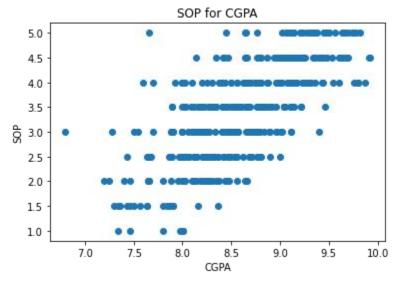
Data Visualization

In [19]:

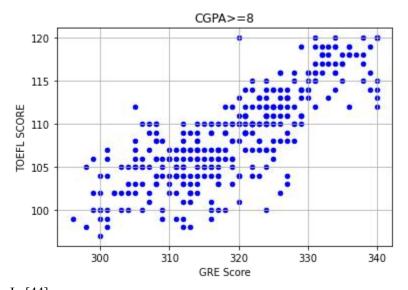
plt.scatter(data['GRE Score'],data['CGPA'])plt.title('CGPA vs GRE Score')plt.xlabel('GRE Score')plt.ylabel('CGPA')plt.show()



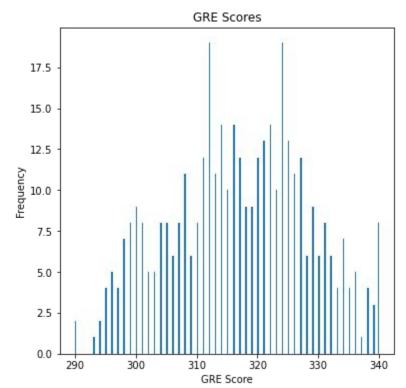
In [42]:
plt.scatter(data['CGPA'],data['SOP'])plt.title('SOP for CGPA')plt.xlabel('CGPA')plt.ylabel('SOP')plt.show()



In [43]:
data[data.CGPA >= 8].plot(kind='scatter', x='GRE Score', y='TOEFL Score',color="BLUE")plt.xlabel("GRE Score")plt.ylabel("TOEFL SCORE")plt.title("CGPA>=8")plt.grid(**True**)plt.show()

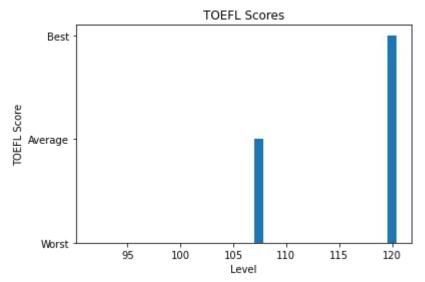


In [44]:
data["GRE Score"].plot(kind = 'hist',bins = 200,figsize = (6,6))plt.title("GRE Scores")plt.xlabel("GRE Score")plt.ylabel("Frequency")plt.show()



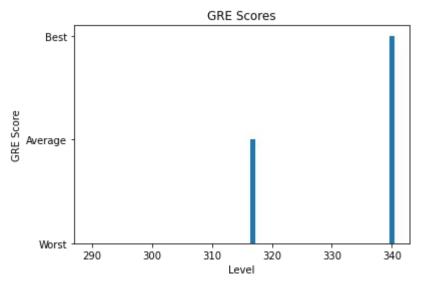
In [45]:

p = np.array([data["TOEFL Score"].min(),data["TOEFL Score"].mean(),data["TOEFL Score"].max()])r = ["Worst","Average","Best"]plt.bar(p,r)plt.title("TOEFL Scores")plt.xlabel("Level")plt.ylabel("TOEFL Score")plt.show()



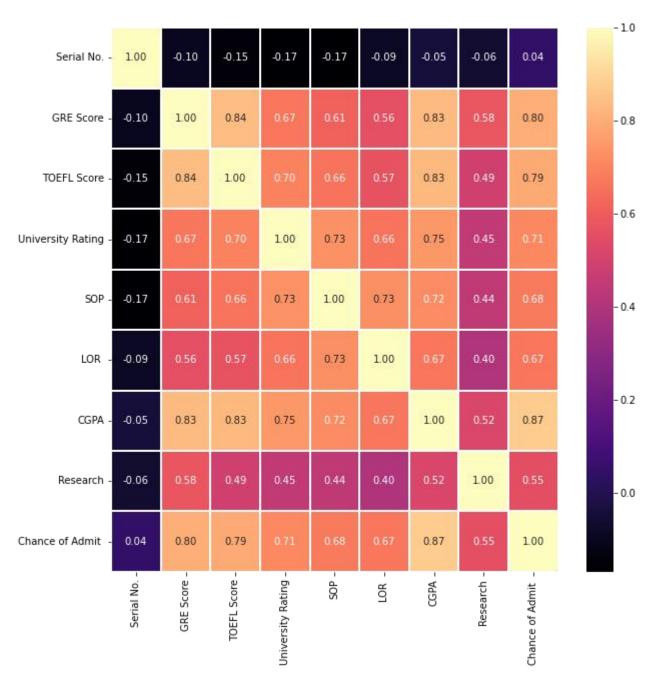
In [46]:

g = np.array([data["GRE Score"].min(),data["GRE Score"].mean(),data["GRE Score"].max()])h = ["Worst","Average","Best"]plt.bar(g,h)plt.title("GRE Scores")plt.xlabel("Level")plt.ylabel("GRE Score")plt.show()

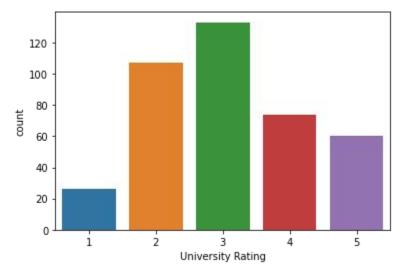


In [47]:

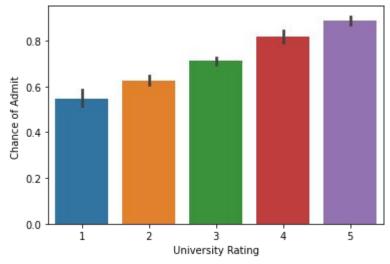
plt.figure(figsize=(10, 10))sns.heatmap(data.corr(), annot=**True**, linewidths=0.05, fmt='.2f',cmap="magma")plt.show()



In [48]: data.Research.value_counts()sns.countplot(x="University Rating",data=data) Out[48]:



In [49]:
sns.barplot(x="University Rating", y="Chance of Admit ", data=data)
Out[49]:



Splitting Dependent and Independent Columns

In [20]:

x = data.iloc[:, :-1]y = data['Chance of Admit']

In [21]:

x.head()

Out[21]:

	Serial No.	GRE Score	TOEFL Score	University Rating	SOP	LOR	CGPA	Research
0	1	337	118	4	4.5	4.5	9.65	1
1	2	324	107	4	4.0	4.5	8.87	1
2	3	316	104	3	3.0	3.5	8.00	1
3	4	322	110	3	3.5	2.5	8.67	1

```
Serial No. GRE Score TOEFL Score University Rating SOP LOR CGPA Research
4
                  314
                               103
                                                     2.0
                                                           3.0
                                                                 8.21
In [22]:
y.head()
Out[22]:
0 0.92
  0.76
1
  0.72
3 0.80
4 0.65
```

Name: Chance of Admit, dtype: float64

Splitting the Data into Train and Test

```
In [23]:

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=50)
```

Model Building

```
In [24]:
from sklearn.ensemble import GradientBoostingRegressormodel =
GradientBoostingRegressor()model.fit(x train,y train)
Out[24]:
GradientBoostingRegressor()
In [25]:
model.score(x_test,y_test)
Out[25]:
0.8045553566319525
In [26]:
from sklearn.metrics import accuracy scorey predict=model.predict(x test)
y train = (y \text{ train}>0.5)y \text{ test} = (y \text{ test}>0.5)
In [28]:
from sklearn.linear model. logistic import LogisticRegressionlore =
LogisticRegression(random state=0, max iter=1000)lr model = lore.fit(x train,
y train)y pred = lr model.predict(x test)
```

Model Evaluation

In [29]:

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))

Accuracy Score: 0.9375

Recall Score: 0.9487179487179487 ROC AUC Score: 0.7243589743589743

Confussion Matrix:

[[1 1] [4 74]]

Save the Model

In [50]:

import pickle

pickle.dump(lr, open("university.pkl", 'wb'))# model = pickle.load(open("university.pkl", 'rb'))

IBM Deployement

In [30]:

!pip install -U ibm-watson-machine-learning

Requirement already satisfied: ibm-watson-machine-learning in /opt/conda/envs/Python-3.9/lib/p ython3.9/site-packages (1.0.257)

Requirement already satisfied: certifi in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (2022.9.24)

Requirement already satisfied: importlib-metadata in /opt/conda/envs/Python-3.9/lib/python3.9/s ite-packages (from ibm-watson-machine-learning) (4.8.2)

Requirement already satisfied: ibm-cos-sdk==2.11.* in /opt/conda/envs/Python-3.9/lib/python3.9 /site-packages (from ibm-watson-machine-learning) (2.11.0)

Requirement already satisfied: lomond in /opt/conda/envs/Python-3.9/lib/python3.9/site-package s (from ibm-watson-machine-learning) (0.3.3)

Requirement already satisfied: pandas<1.5.0,>=0.24.2 in /opt/conda/envs/Python-3.9/lib/python3. 9/site-packages (from ibm-watson-machine-learning) (1.3.4)

Requirement already satisfied: tabulate in /opt/conda/envs/Python-3.9/lib/python3.9/site-package s (from ibm-watson-machine-learning) (0.8.9)

Requirement already satisfied: packaging in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (21.3)

Requirement already satisfied: urllib3 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (1.26.7)

Requirement already satisfied: requests in /opt/conda/envs/Python-3.9/lib/python3.9/site-package s (from ibm-watson-machine-learning) (2.26.0)

Requirement already satisfied: ibm-cos-sdk-core=2.11.0 in /opt/conda/envs/Python-3.9/lib/pyth on3.9/site-packages (from ibm-cos-sdk=2.11.*->ibm-watson-machine-learning) (2.11.0)

Requirement already satisfied: ibm-cos-sdk-s3transfer==2.11.0 in /opt/conda/envs/Python-3.9/lib /python3.9/site-packages (from ibm-cos-sdk==2.11.*->ibm-watson-machine-learning) (2.11.0)

Requirement already satisfied: jmespath<1.0.0,>=0.7.1 in /opt/conda/envs/Python-3.9/lib/python 3.9/site-packages (from ibm-cos-sdk==2.11.*->ibm-watson-machine-learning) (0.10.0)

Requirement already satisfied: python-dateutil<3.0.0,>=2.1 in /opt/conda/envs/Python-3.9/lib/pyt hon3.9/site-packages (from ibm-cos-sdk-core==2.11.0->ibm-cos-sdk==2.11.*->ibm-watson-mac hine-learning) (2.8.2)

Requirement already satisfied: pytz>=2017.3 in /opt/conda/envs/Python-3.9/lib/python3.9/site-pa ckages (from pandas<1.5.0,>=0.24.2->ibm-watson-machine-learning) (2021.3)

Requirement already satisfied: numpy>=1.17.3 in /opt/conda/envs/Python-3.9/lib/python3.9/sitepackages (from pandas<1.5.0,>=0.24.2->ibm-watson-machine-learning) (1.20.3)

Requirement already satisfied: six>=1.5 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packag es (from python-dateutil<3.0.0,>=2.1->ibm-cos-sdk-core==2.11.0->ibm-cos-sdk==2.11.*->ibmwatson-machine-learning) (1.15.0)

Requirement already satisfied: idna<4,>=2.5 in /opt/conda/envs/Python-3.9/lib/python3.9/site-pa ckages (from requests->ibm-watson-machine-learning) (3.3)

Requirement already satisfied: charset-normalizer~=2.0.0 in /opt/conda/envs/Python-3.9/lib/pyth on 3.9/site-packages (from requests->ibm-watson-machine-learning) (2.0.4)

Requirement already satisfied: zipp>=0.5 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packa ges (from importlib-metadata->ibm-watson-machine-learning) (3.6.0)

Requirement already satisfied: pyparsing!=3.0.5,>=2.0.2 in /opt/conda/envs/Python-3.9/lib/pytho n3.9/site-packages (from packaging->ibm-watson-machine-learning) (3.0.4)

```
In [31]:
```

Out[34]: 'SUCCESS'

```
from ibm watson machine learning import APIClientimport json
In [32]:
wml credentials = {
  "apikey":"2P76TOf0kpxjsnHwF-xwPOlMqVBBviAH0TPSUWUnUb4M",
  "url":"https://us-south.ml.cloud.ibm.com"}
In [33]:
wml client = APIClient(wml credentials)wml client.spaces.list()
Note: 'limit' is not provided. Only first 50 records will be displayed if the number of records exce
ed 50
ID
                     NAME CREATED
labc7d7a-f529-4419-9afd-b5de22df88be models 2022-11-16T11:23:56.373Z
______
In [34]:
SPACE ID = "1abc7d7a-f529-4419-9afd-
b5de22df88be"wml client.set.default space(SPACE ID)
```

NAME ASSET ID **TYPE** default py3.6 0062b8c9-8b7d-44a0-a9b9-46c416adcbd9 base kernel-spark3.2-scala2.12 020d69ce-7ac1-5e68-ac1a-31189867356a base pytorch-onnx 1.3-py3.7-edt 069ea134-3346-5748-b513-49120e15d288 base scikit-learn 0.20-py3.6 09c5a1d0-9c1e-4473-a344-eb7b665ff687 base spark-mllib_3.0-scala_2.12 09f4cff0-90a7-5899-b9ed-1ef348aebdee base pytorch-onnx rt22.1-py3.9 0b848dd4-e681-5599-be41-b5f6fccc6471 base ai-function 0.1-py3.6 0cdb0f1e-5376-4f4d-92dd-da3b69aa9bda base shiny-r3.6 0e6e79df-875e-4f24-8ae9-62dcc2148306 base 1092590a-307d-563d-9b62-4eb7d64b3f22 base tensorflow 2.4-py3.7-horovod pytorch 1.1-py3.6 10ac12d6-6b30-4ccd-8392-3e922c096a92 base tensorflow 1.15-py3.6-ddl 111e41b3-de2d-5422-a4d6-bf776828c4b7 base autoai-kb rt22.2-py3.10 125b6d9a-5b1f-5e8d-972a-b251688ccf40 base runtime-22.1-py3.9 12b83a17-24d8-5082-900f-0ab31fbfd3cb base scikit-learn 0.22-py3.6 154010fa-5b3b-4ac1-82af-4d5ee5abbc85 base default r3.6 1b70aec3-ab34-4b87-8aa0-a4a3c8296a36 base pytorch-onnx 1.3-py3.6 1bc6029a-cc97-56da-b8e0-39c3880dbbe7 base kernel-spark3.3-r3.6 1c9e5454-f216-59dd-a20e-474a5cdf5988 base pytorch-onnx rt22.1-py3.9-edt 1d362186-7ad5-5b59-8b6c-9d0880bde37f base tensorflow 2.1-py3.6 1eb25b84-d6ed-5dde-b6a5-3fbdf1665666 base spark-mllib 3.2 20047f72-0a98-58c7-9ff5-a77b012eb8f5 base tensorflow 2.4-py3.8-horovod 217c16f6-178f-56bf-824a-b19f20564c49 base runtime-22.1-py3.9-cuda 26215f05-08c3-5a41-a1b0-da66306ce658 base do py3.8 295addb5-9ef9-547e-9bf4-92ae3563e720 base 2aa0c932-798f-5ae9-abd6-15e0c2402fb5 base autoai-ts_3.8-py3.8 tensorflow 1.15-py3.6 2b73a275-7cbf-420b-a912-eae7f436e0bc base kernel-spark3.3-py3.9 2b7961e2-e3b1-5a8c-a491-482c8368839a base 2c8ef57d-2687-4b7d-acce-01f94976dac1 base pytorch 1.2-py3.6 spark-mllib 2.3 2e51f700-bca0-4b0d-88dc-5c6791338875 base pytorch-onnx 1.1-py3.6-edt 32983cea-3f32-4400-8965-dde874a8d67e base spark-mllib 3.0-py37 36507ebe-8770-55ba-ab2a-eafe787600e9 base spark-mllib 2.4 390d21f8-e58b-4fac-9c55-d7ceda621326 base autoai-ts rt22.2-py3.10 396b2e83-0953-5b86-9a55-7ce1628a406f base xgboost 0.82-py3.6 39e31acd-5f30-41dc-ae44-60233c80306e base 40589d0e-7019-4e28-8daa-fb03b6f4fe12 base pytorch-onnx 1.2-py3.6-edt

40 7207 702 7525 126 0 0104201421 1
pytorch-onnx_rt22.2-py3.10
default_r36py38 41c247d3-45f8-5a71-b065-8580229facf0 base
autoai-ts_rt22.1-py3.9 4269d26e-07ba-5d40-8f66-2d495b0c71f7 base
autoai-obm_3.0 42b92e18-d9ab-567f-988a-4240ba1ed5f7 base
pmml-3.0_4.3 493bcb95-16f1-5bc5-bee8-81b8af80e9c7 base
spark-mllib_2.4-r_3.6 49403dff-92e9-4c87-a3d7-a42d0021c095 base
xgboost_0.90-py3.6 4ff8d6c2-1343-4c18-85e1-689c965304d3 base
pytorch-onnx_1.1-py3.6 50f95b2a-bc16-43bb-bc94-b0bed208c60b base
autoai-ts_3.9-py3.8 52c57136-80fa-572e-8728-a5e7cbb42cde base
spark-mllib_2.4-scala_2.11 55a70f99-7320-4be5-9fb9-9edb5a443af5 base
spark-mllib_3.0 5c1b0ca2-4977-5c2e-9439-ffd44ea8ffe9 base
autoai-obm_2.0 5c2e37fa-80b8-5e77-840f-d912469614ee base
spss-modeler_18.1 5c3cad7e-507f-4b2a-a9a3-ab53a21dee8b base
cuda-py3.8 5d3232bf-c86b-5df4-a2cd-7bb870a1cd4e base
autoai-kb_3.1-py3.7 632d4b22-10aa-5180-88f0-f52dfb6444d7 base
pytorch-onnx_1.7-py3.8 634d3cdc-b562-5bf9-a2d4-ea90a478456b base
spark-mllib_2.3-r_3.6 6586b9e3-ccd6-4f92-900f-0f8cb2bd6f0c base
tensorflow_2.4-py3.7 65e171d7-72d1-55d9-8ebb-f813d620c9bb base
spss-modeler_18.2 687eddc9-028a-4117-b9dd-e57b36f1efa5 base
pytorch-onnx_1.2-py3.6 692a6a4d-2c4d-45ff-a1ed-b167ee55469a base
spark-mllib_2.3-scala_2.11 7963efe5-bbec-417e-92cf-0574e21b4e8d base
spark-mllib_2.4-py37
caffe_1.0-py3.6 7bb3dbe2-da6e-4145-918d-b6d84aa93b6b base
pytorch-onnx_1.7-py3.7 812c6631-42b7-5613-982b-02098e6c909c base
cuda-py3.6 82c79ece-4d12-40e6-8787-a7b9e0f62770 base
tensorflow_1.15-py3.6-horovod 8964680e-d5e4-5bb8-919b-8342c6c0dfd8 base
hybrid 0.1 8c1a58c6-62b5-4dc4-987a-df751c2756b6 base
pytorch-onnx_1.3-py3.7 8d5d8a87-a912-54cf-81ec-3914adaa988d base
caffe-ibm 1.0-py3.6 8d863266-7927-4d1e-97d7-56a7f4c0a19b base
spss-modeler 17.1 902d0051-84bd-4af6-ab6b-8f6aa6fdeabb base
do 12.10 9100fd72-8159-4eb9-8a0b-a87e12eefa36 base
do py3.7 9447fa8b-2051-4d24-9eef-5acb0e3c59f8 base
spark-mllib_3.0-r_3.6 94bb6052-c837-589d-83f1-f4142f219e32 base
cuda-py3.7-opence 94e9652b-7f2d-59d5-ba5a-23a414ea488f base
nlp-py3.8 96e60351-99d4-5a1c-9cc0-473ac1b5a864 base
cuda-py3.7 9a44990c-1aa1-4c7d-baf8-c4099011741c base
hybrid 0.2 9b3f9040-9cee-4ead-8d7a-780600f542f7 base
spark-mllib 3.0-py38 9f7a8fc1-4d3c-5e65-ab90-41fa8de2d418 base
-r pjes yi/woiei iwaa aada wayo iiiwowaa iio owso

```
a545cca3-02df-5c61-9e88-998b09dc79af base
autoai-kb 3.3-py3.7
                          a6082a27-5acc-5163-b02c-6b96916eb5e0 base
spark-mllib 3.0-py39
runtime-22.1-py3.9-do
                           a7e7dbf1-1d03-5544-994d-e5ec845ce99a base
default py3.8
                       ab9e1b80-f2ce-592c-a7d2-4f2344f77194 base
tensorflow rt22.1-py3.9
                           acd9c798-6974-5d2f-a657-ce06e986df4d base
kernel-spark3.2-py3.9
                          ad7033ee-794e-58cf-812e-a95f4b64b207 base
autoai-obm 2.0 with Spark 3.0 af10f35f-69fa-5d66-9bf5-acb58434263a base
default py3.7 opence
                           c2057dd4-f42c-5f77-a02f-72bdbd3282c9 base
tensorflow 2.1-py3.7
                          c4032338-2a40-500a-beef-b01ab2667e27 base
do py3.7 opence
                         cc8f8976-b74a-551a-bb66-6377f8d865b4 base
spark-mllib 3.3
                        d11f2434-4fc7-58b7-8a62-755da64fdaf8 base
                         d139f196-e04b-5d8b-9140-9a10ca1fa91a base
autoai-kb 3.0-py3.6
                          d82546d5-dd78-5fbb-9131-2ec309bc56ed base
spark-mllib_3.0-py36
autoai-kb 3.4-py3.8
                         da9b39c3-758c-5a4f-9cfd-457dd4d8c395 base
kernel-spark3.2-r3.6
                         db2fe4d6-d641-5d05-9972-73c654c60e0a base
autoai-kb rt22.1-py3.9
                          db6afe93-665f-5910-b117-d879897404d9 base
tensorflow rt22.1-py3.9-horovod dda170cc-ca67-5da7-9b7a-cf84c6987fae base
autoai-ts 1.0-py3.7
                         deef04f0-0c42-5147-9711-89f9904299db base
tensorflow 2.1-py3.7-horovod
                              e384fce5-fdd1-53f8-bc71-11326c9c635f base
                       e4429883-c883-42b6-87a8-f419d64088cd base
default py3.7
do 22.1
                     e51999ba-6452-5f1f-8287-17228b88b652 base
autoai-obm 3.2
                        eae86aab-da30-5229-a6a6-1d0d4e368983 base
                           f65bd165-f057-55de-b5cb-f97cf2c0f393 base
tensorflow rt22.2-py3.10
do 20.1
                     f686cdd9-7904-5f9d-a732-01b0d6b10dc5 base
pytorch-onnx_rt22.2-py3.10-edt f8a05d07-e7cd-57bb-a10b-23f1d4b837ac base
scikit-learn 0.19-py3.6
                          f963fa9d-4bb7-5652-9c5d-8d9289ef6ad9 base
tensorflow 2.4-py3.8
                          fe185c44-9a99-5425-986b-59bd1d2eda46 base
In [36]:
import sklearnsklearn. version
Out[36]:
'1.0.2'
In [37]:
MODEL NAME = 'univ ad'
DEPLOYMENT NAME = 'iris adm'
DEMO MODEL = lr model
In [38]:
software spec uid = wml client.software specifications.get id by name('runtime-22.1-py3.9')
```

```
model props = {
   wml client.repository.ModelMetaNames.NAME: MODEL NAME,
   wml client.repository.ModelMetaNames.TYPE: 'scikit-learn 1.0',
   wml client.repository.ModelMetaNames.SOFTWARE SPEC UID: software spec uid}
model details = wml client.repository.store model(
   model = DEMO MODEL,
   meta props = model props,
   training data = x train,
   training target = y train)
model details
Out[38]:
{'entity': {'hybrid pipeline software specs': [],
 'label column': 'Chance of Admit',
 'schemas': {'input': [{'fields': [{'name': 'Serial No.', 'type': 'int64'},
   {'name': 'GRE Score', 'type': 'int64'},
   {'name': 'TOEFL Score', 'type': 'int64'},
   {'name': 'University Rating', 'type': 'int64'},
   {'name': 'SOP', 'type': 'float64'},
   {'name': 'LOR ', 'type': 'float64'},
   {'name': 'CGPA', 'type': 'float64'},
   {'name': 'Research', 'type': 'int64'}],
   'id': '1',
   'type': 'struct'}],
 'output': []},
 'software spec': {'id': '12b83a17-24d8-5082-900f-0ab31fbfd3cb',
 'name': 'runtime-22.1-py3.9'},
 'type': 'scikit-learn 1.0'},
'metadata': {'created at': '2022-11-16T17:11:57.291Z',
 'id': '0944c716-c6b4-4221-b6b6-893d1d74c506',
 'modified at': '2022-11-16T17:11:59.723Z',
 'name': 'univ ad',
 'owner': 'IBMid-667000F3HX',
 'resource key': 'cdd58957-c38c-48ca-9af5-e012d9239d60',
 'space id': 'labc7d7a-f529-4419-9afd-b5de22df88be'},
'system': {'warnings': []}}
```

```
In [39]:
model id = wml client.repository.get model id(model details)model id
Out[39]:
'0944c716-c6b4-4221-b6b6-893d1d74c506'
In [40]:
deployment props = {
  wml client.deployments.ConfigurationMetaNames.NAME:DEPLOYMENT NAME,
  wml client.deployments.ConfigurationMetaNames.ONLINE: {}}
deployment = wml client.deployments.create(
  artifact uid = model id,
  meta props = deployment props)
#########
Synchronous deployment creation for uid: '0944c716-c6b4-4221-b6b6-893d1d74c506' started
##########
initializing
Note: online url is deprecated and will be removed in a future release. Use serving urls instead.
ready
Successfully finished deployment creation, deployment uid='363f2f17-a1e5-4b6b-93d1-a79ae7b
04882'
```

HTML CODES

Uploading Html codes

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>
```

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-</pre>
scale=1.0">
  <link rel="stylesheet" href="/css/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"</pre>
required>
    </div>
    <div class="form-data">
      <label for="univRank">University Rank:</label>
      <input name="univRank" placeholder="University Rank"</pre>
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>
<br>
<br/>
<button type="submit" class="btns">Submit</button></a href="/"><button type="button"</td>

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

Index.html

```
<br/>
```

Nochance.html

```
<h1>University Admit Eligibility Predictor</h1>
<a href="/checkEligibility"><button type="button" class="btns">Check Eligibility</button></a>
</div>
</body>
</html>
```

Script.js

```
const button =
document.getElementById('button');
const theForm =
document.getElementById('theForm');
const loading =
document.getElementById('spinner');

const disableButton = () => {
   console.log('Submitting form...');
   button.disabled = true;
   button.className = "btn btn-outline-primary";
   button.innerHTML = "Predicting..."
   loading.style.display = "block"
};
```

```
const enableButton = () => {
  console.log('Loading window...');
  button.disabled = false;
```

```
button.className = "btn btn-primary"
button.innerHTML = "Predict"
loading.style.display = "none"
}
theForm.onsubmit = disableButton;
window.onload = enableButton;
```

Styles.css

```
*{
    margin: 0;
    box-sizing: border-box;
}
.index-page{
    background-image: url(/img/bg.jpeg);
}
.index{
    display: flex;
```

```
flex-direction: column;
  width: fit-content;
  height: 20%;
  top: 50%;
  left: 50%;
  transform: translate(-50%, -50%);
  position: absolute;
  padding: 20px;
  border-radius: 50px;
  background: rgba(255, 255, 255, 0.7);
  align-items: center;
  justify-content: center;
  box-shadow: 2px 2px 20px rgb(0, 0, 0);
}
.home {
  background-image: url(/img/bg2.jpeg);
}
.home h1 {
  width: 100%;
  text-align: center;
  margin-top: 5px;
}
.form{
  display: flex;
```

```
top: 50%;
  left: 50%;
  position: absolute;
  transform: translate(-50%, -50%);
  flex-direction: column;
  width: fit-content;
  height: fit-content;
  padding: 20px;
  border-radius: 50px;
  background: rgba(255, 255, 255, 0.8);
  align-items: center;
  justify-content: center;
  box-shadow: 2px 2px 20px rgb(0, 0, 0);
}
.form input{
  width: 400px;
  height: 50px;
  margin-right: 10px;
  margin-block: 20px;
  font-size: large;
  border-radius: 20px;
  border: none;
  padding: 12px;
  box-shadow: 1px 1px 5px rgb(0, 0, 0);
  background-color: rgba(255, 255, 255, 0.8);
  color: #000;
```

```
}
. form\text{-}data \{
  display: flex;
  flex-direction: row;
}
. form\text{-}data\ label \{
  width: 150px;
  height: 50px;
  margin-left: 10px;
  margin-block: 20px;
  align-items: center;
  justify-content: center;
  font-size: large;
  padding: 12px;
  color: #000;
}
.predict{
  background-image: url(/img/bg1.jpeg);
  background-size: cover;
}
.prediction-box{
  background-color: #fff;
  width: fit-content;
```

```
height: fit-content;
  padding: 20px;
  display: flex;
  flex-direction: column;
  top: 50%;
  left: 30%;
  position: absolute;
  transform: translate(-50%, -30%);
  border-radius: 30px;
  background: rgba(255, 255, 255, 0.9);
  box-shadow: 2px 2px 20px rgb(0, 0, 0);
}
.prediction \{\\
  display: flex;
  flex-direction: row;
  align-items: center;
  padding: 20px;
  margin-block: 20px;
}
.prediction h3{
  text-decoration: underline;
}
.btns{
  width: 200px;
```

```
height: 50px;
font-size: larger;
font-weight: 500;
border-radius: 20px;
border: none;
margin-top: 20px;
background-color: #fff;
transition: 200ms ease-in-out;
cursor: pointer;
}

.btns:hover{
  background-color: #000;
  color: #fff;
  transform: scale(1.1);
}
```

PYTHON CODE

App.py

```
from flask import Flask, render_template, request
import requests
```

```
API KEY =
"dAkQTmsJ7sfRzutZ8fTcNbHZvKD ZyoxqjtYF7h8VwC7"
token response =
requests.post('https://iam.cloud.ibm.com/identity/token', data={"apikey":
API_KEY, "grant_type": 'urn:ibm:params:oauth:grant-type:apikey'})
mltoken = token response.json()["access token"]
header = {'Content-Type': 'application/json', 'Authorization': 'Bearer ' +
mltoken}
app = Flask(__name__, static_url_path=")
@app.route('/')
def index():
  return render_template('index.html')
@app.route('/checkEligibility')
```

```
def checkEligibility():
  return render template('Demo2.html')
@app.route('/predict', methods=['POST'])
def predict():
  greScore = int(request.form['greScore'])
  toeflScore = int(request.form['toeflScore'])
  univRank = int(request.form['univRank'])
  sop = float(request.form['sop'])
  lor = float(request.form['lor'])
  cgpa = float(request.form['cgpa'])
  research = int(request.form['research'])
  array of input fields = ['greScore', 'toeflScore', 'univRank', 'sop', 'lor',
'cgpa', 'research']
  array of values to be scored = [greScore, toeflScore, univRank, sop,
lor, cgpa, research]
  payload scoring = {"input data": [{"fields": [array of input fields],
"values": [array of values to be scored]}]}
  response scoring = requests.post('https://us-
south.ml.cloud.ibm.com/ml/v4/deployments/9f4939ed-7f21-4881-8ae4-
234e7515f65a/predictions?version=2022-10-21', json=payload scoring,
headers={'Authorization': 'Bearer ' + mltoken})
  predictions = response scoring.json()
```

```
prediction = predictions['predictions'][0]['values'][0][0]

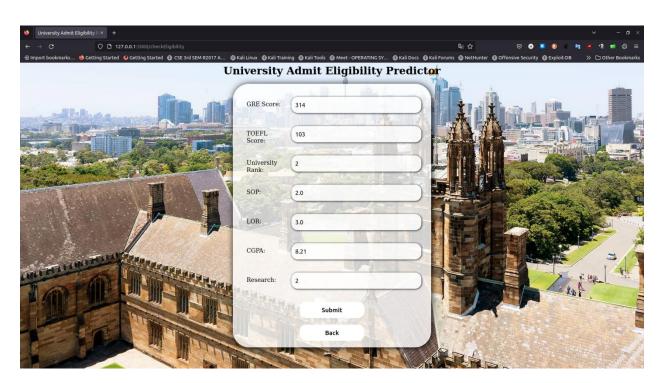
if prediction:
    return render_template('chance.html')

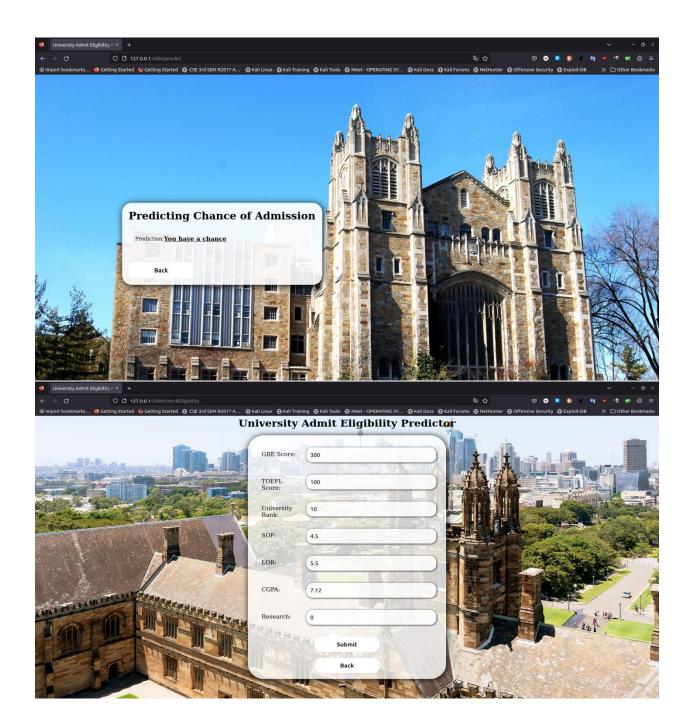
else:
    return render_template('noChance.html')

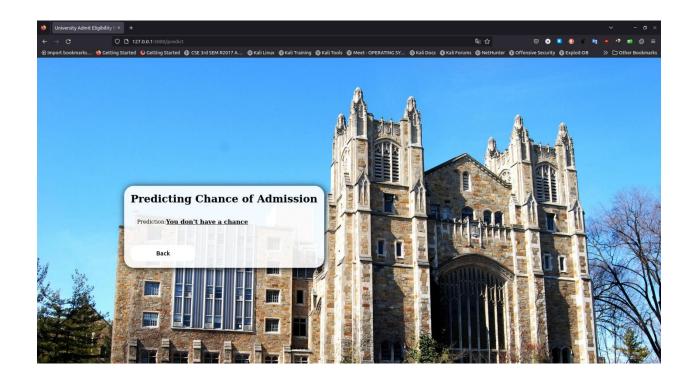
if __name__ == "__main__":
```

OUTPUT IMAGES:

app.run()







13.2 GITHUB LINK:

https://github.com/IBM-EPBL/IBM-Project-8830-1658933471

13.3 DEMO LINK:

https://drive.google.com/file/d/1ypUkgaxFIdYUEm-y5tKUFOKNWMoVABPb/view?usp=sharing