

IBM-Project-21048-1659771001

Project Name: University Admit Eligibility Predictor

Team Id : PNT2022TMID20129

Domain : Applied Data Science

Team Members:

Dhanarajan G (Team Leader)

Gokul Karthick P

Sanjay Kumar S

Ajay Rajendran S

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INTRODUCTION

1.1 Project overview

Students are often worried about their chances of admission to university. The aim of this project is to help students in shortlisting universities with their profiles. The predicted output gives them a fair idea about their admission chances in a particular university. This analysis should also help students who are currently preparing or will be preparing to get better idea.

1.2 Purpose

This university admit eligibility predictor is a web based application in which students can register with their personal as well as mark details for prediction the colleges and the administrator can allot the seats for the students. Administrator can add the college details and he batch details. Using this software, the entrance seat allotment became easier and can be implemented in the system. The main advantage of the project is the computerization of the entrance seat allotment process. Administrator has the power of the allotment. He can add the allotted seats into a file and the details are saved into a file and the details are saved into the system. The total time for the entrance allotment became lesser and the allotment process became faster. It helps student for making decision for choosing a right college and money that they have to spend at education consultancy firms and also it will help them to limit their

application to small number by proving them the suggestion of the universities where they have best chance of securing admission thus saving more money on the application fees.

LITERATURE SURVEY

2.1 Existing problem

In this web based application, the scope of this project is a web application that allows user to enter their academic data and get predictions of their chances of admissions in the university tier of their choosing. It also provides them to answer to most common FAQ's that arise when thinking of admissions for post graduate studies. It also provides analysis based on the dataset shows how different parameters affect the chances of admissions. A Database will also be implemented for the system so that students can save their profile. Issues of web security other than password protection within the website are part of this project. Another issue is to avoid time consuming high and make the task completed the task in a quickly manner.

2.2 References

- [1] Abdul Fatah S; M, A. H. (2012). Hybrid Recommender System for Predicting College Admission, pp. 107–113.
- [2] Bibodi, J., Vadodaria, A., Rawat, A. and Patel, J. (n.d.). Admission Prediction System Using Machine Learning.
- [3] Jamison, Applying Machine learning to predict Davidson college's admissions Yield , pp 765-766(2017)
- [4] Eberle, W., Simpson, E., Talbert, D., Roberts, L. and Pope, A. (n.d.). Using Machine Learning and Predictive Modeling to Assess Admission Policies and Standards.
- [5] Lapovsky, L The Changing Business for Colleges and Universities. Forbes(2018)
- [6] Mane, R. V. (2016). Predicting Student Admission decisions by Association Rule Mining with Pattern Growth Approach, pp. 202–207
- [7] J. Bodailla et al. "Knowledge -Based System" Elsevier B.V
- [8] B. Ghai , "Analysis and prediction of american graduate admissions Process"2018
- [9] B.K. Bardwaj and s. pal "mining education data to analyse student's performance",2012
- [10] c.c. Aggarwal, Data mining: The textbook ,2015

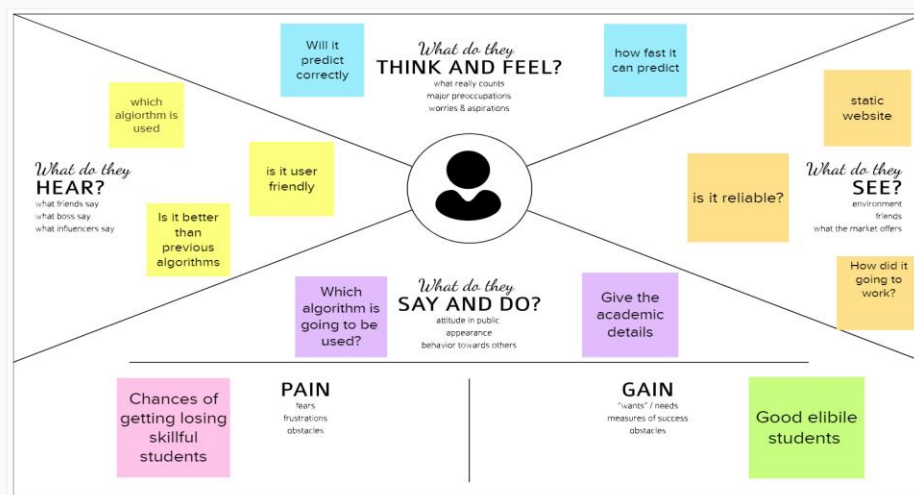
2.3 Problem Statement Definition

The problem in this process it is far away from the place so it takes time to for apply college and process is very tedious. It is also very difficult when it comes to offline and it takes long time for the result of the application. Because of time consumption and delay in result makes student's life questionable and career path unclear. This will be pioneer to make students to choose correct career and

IDEATION & PROPOSED SOLUTION

3.1 Empathy Map Canvas

An empathy map is a collaborative tool teams can use to gain a deeper insight into their customers. Much like a user persona, an empathy map can represent a group of users, such as a customer segment. The empathy map was originally created by Dave Gray and has gained much popularity within the agile community.



Dhanarajan G (1912055)
 Gokul Karthick P (1912057)
 Sanjay Kumar S (1912045)
 Ajay Ranjendran S (1912002)

3.2 Ideation & Brainstorming



Ideation is often closely related to the practice of brainstorming, a specific technique that is utilized to generate new ideas. A principal difference between ideation and brainstorming is that ideation is commonly more thought of as being an individual pursuit, while brainstorming is almost always a group activity. As you can see, ideation is not just a one time idea generation or a brainstorming session. In fact, we can divide ideation in these three stages: generation, selection, and development. Brainstorming is one of the primary methods employed during the Ideation stage of a typical Design Thinking process.

Brainstorming is a method of generating ideas and sharing knowledge to solve a particular commercial or technical problem, in which participants are encouraged to think without interruption. Brainstorming is a group activity where each participant shares their ideas as soon as they come to mind. Group brainstorming stimulates creativity and invites participation from everyone, making it a great tool for generating a wide variety of ideas in a short amount of time. It's especially helpful when trying to solve a problem that you are really close to. Sometimes getting outside perspectives can breathe new life into a project and drive momentum towards a solution.

3.3 Problem Solution

Problem solving is a basic task for the project management. It is a process for developing and applying a solution for the occurred problems. The probability of the success rises, if a particular method is implemented to the project work.

3.4 Problem Solution fit

The Problem-Solution Fit simply means that you have found a problem with your customer and that the solution you have realized for it actually solves the customer's problem.

1. Customer Segment

The possible student who have completed their schooling and UG searching for university to study PG

2. Jobs to be done

The major task is to design a university admission prediction system and to provide probabilistic insight into the university rating, cutoffs, intake count

3. Triggers

Students often get tensed and anxious about their admission chances of their desired universities

4. Emotions

Before : Insecure and unaware of the process, suffering to select the best suited-university

After : Secure, user friendly and aware of process

5. Available Solutions

Lack dynamic nature and scalability.
Incomplete training information.

REQUIREMENT ANALYSIS

4.1 Functional requirements

A functional requirement defines a function of a system or its component, where a function is described as a specification of behavior between inputs and outputs. Functional requirements may involve calculations, technical details, data manipulation and processing, and other specific functionality that define what a system is supposed to accomplish. Behavioral requirements describe all the cases where the system uses the functional requirements, these are captured in use cases.

Following are the functional requirements of the proposed solution.

FR No.	Functional Requirement(Epic)	Sub Requirement (Story / Sub-Task)
FR-1	Registration of User	Registration via forms Registration via gmail
FR-2	Confirmation of User	Confirmation via Email Confirmation via OTP
FR-3	User Details	Submit the documents <ul style="list-style-type: none">GRE/TOEFL mark sheetResume/BioRecommendation Letter
FR-4	User Requirements	<ul style="list-style-type: none">Have to upload the required relevant documents in the specified location in the websiteAfter Observing the uploads, the system would scrape all the required information for predictionList all possible university for the student shown based on the collected information

4.2 Non Functional requirements

Nonfunctional Requirements (NFRs) define system attributes such as security, reliability, performance, maintainability, scalability, and usability. They serve as constraints or restrictions on the design of the system across the different backlogs.

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

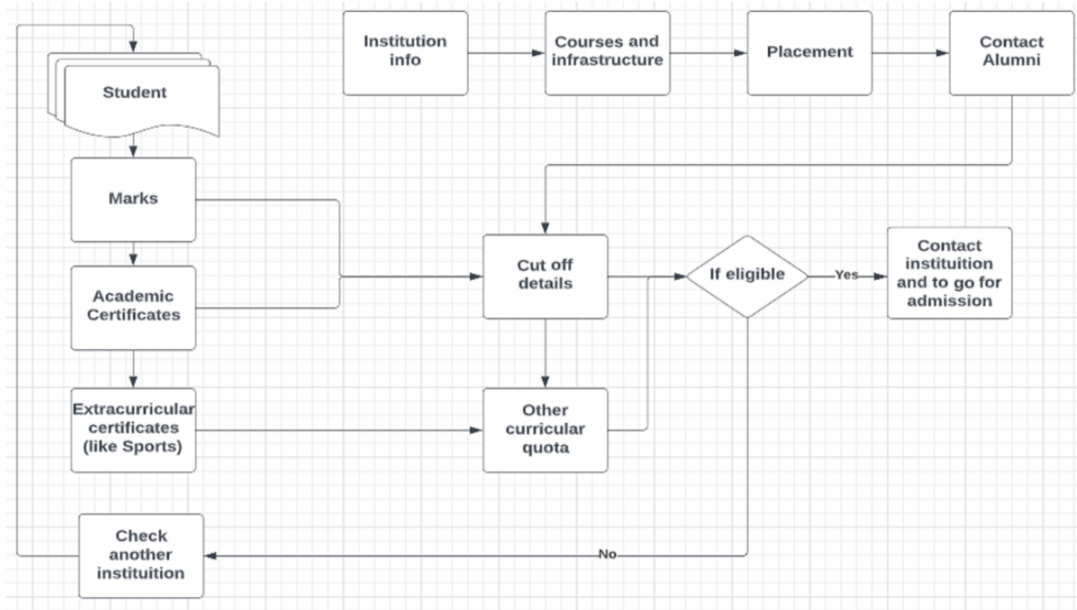
FR No.	Non-Functional Requirement	Description
NFR-1	Usability	<ul style="list-style-type: none">• This system doesn't expect any technical pre-requisite/knowledge from the user• It is very user friendly• Reduced focus on Short Term memoryload Focus on Internal Locus of Control• It load the content and display them (< 30 seconds) i.e user doesn't wait for a while• The fields in the website is quite easy to understand and self explanatory
NFR-2	Security	<ul style="list-style-type: none">• The authenticated user could be able to access the services of the site.• User Data is saved and updated every hour

		<ul style="list-style-type: none">• In case of failure, the system should be able to come back to normal operation with in 1 hour
NFR-3	Reliability	<ul style="list-style-type: none">• The website would always try to get back normal with maximum reliability due to the damages that could be caused because of incomplete and incorrect data
NFR-4	Performance	<ul style="list-style-type: none">• It can balance the traffic efficiently by service the request quickly as much as possible
NFR-5	Availability	<ul style="list-style-type: none">• Little bit of data redundancy• There may be some prone to error• Quick,Compatible and robust• It works 24x7
NFR-6	Scalability	<ul style="list-style-type: none">• The number of students is getting increasing now days so it is crucial that a more number of users/students should be able to access the system at the same time without any delay• The admission season is probably when the system will be under the moststrain.• It must therefore be able to manage numerous concurrent users.

PROJECT DESIGN

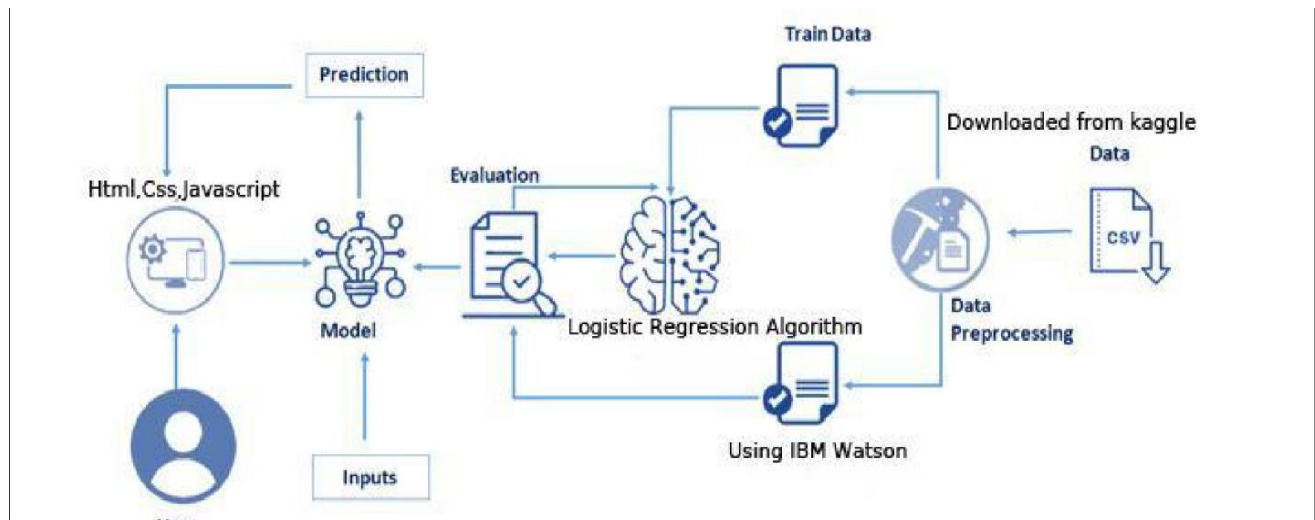
5.1 Data Flow Diagrams

Data flow diagram is a graphical or visual representation using a standardized set of symbols and notations to describe the business operations through data movement.



5.2 Solution and Technical Architecture

Technical Architecture (TA) is a form of IT architecture that is used to design computer systems. It involves the development of a technical blueprint with regard to the arrangement, interaction, and interdependence of all elements so that system relevant requirements are met



5.3 User Stories

A user story is an informal, general explanation of a software feature written from the perspective of the end user or customer. The purpose of a user story is to articulate how a piece of work will deliver a particular value back to the customer

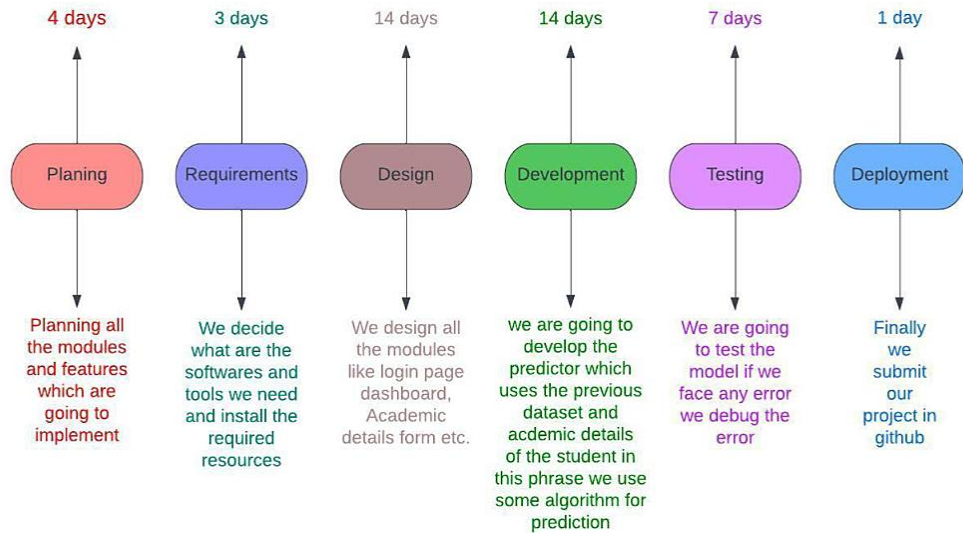
User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance criteria	Priority	Release
Customer	Landing page	USN-1	As a user, I can view the details about the institution.	I can access the university landing page	Medium	Sprint-1
		USN-2	As a user, I can view courses, infrastructure of university.	I can get the details about the university.	Medium	Sprint-1
		USN-3	As a user, I can view the placement and training details.	I can ensure the placement confidence.	Medium	Sprint-1
		USN-4	As a user, I can fill the contact form for queries.	I can fill and submit the contact form	Low	Sprint-2
		USN-5	As a user, I can see the social media (instagram, linkedin, facebook) profiles of the university.	I can reach out to them via social media	Medium	Sprint-1
		USN-6	As a user, I can see testimonials of students who graduated from the university	I can access the testimonials	Medium	Sprint-1
	Admissions	USN-7	As a user, I can see the previous and present year cut-off marks and any extracurricular quota (sports).	I can ensure the marks and quota details. I can also download as a document.	High	Sprint-2
		USN-8	As a user, I can read about proud alumni of the university. If possible, I can contact him/her about this institution.	I can access the details of alumni of the university	Medium	Sprint-2

PROJECT PLANNING AND SCHEDULING

6.1 Sprint Planning & Estimation

University Admit Eligibility Predictor

Milestone and Activity list



6.2 Sprint Delivery Schedule

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	Registration	USN-1	As a user, you can register in the application by entering your email address, password, and confirming the password	2	High	Gokul Karthick P
Sprint-1		USN-2	As a user, you will receive a confirmation email after registering in the application	1	High	Dhanarajan G
Sprint-2		USN-3	As a user, you can register in the application via Facebook	2	Low	Ajay Rajendran S
Sprint-1	Login	USN-4	As a user, you can login to the application by entering your email and password	1	High	Sanjay Kumar S

Sprint	Total Story Points	Duration	Sprint Start Date	Sprint End Date (Planned)	Story Points Completed (as on Planned End Date)	Sprint Release Date (Actual)
Sprint-1	20	5 Days	29 Oct 2022	04 Nov 2022	20	03 Nov 2022
Sprint-2	20	4 Days	04 Oct 2022	08 Nov 2022	20	07 Nov 2022
Sprint-3	20	4 Days	08 Nov 2022	11 Nov 2022	20	10 Nov 2022
Sprint-4	20	4 Days	11 Nov 2022	14 Nov 2022	20	13 Nov 2022

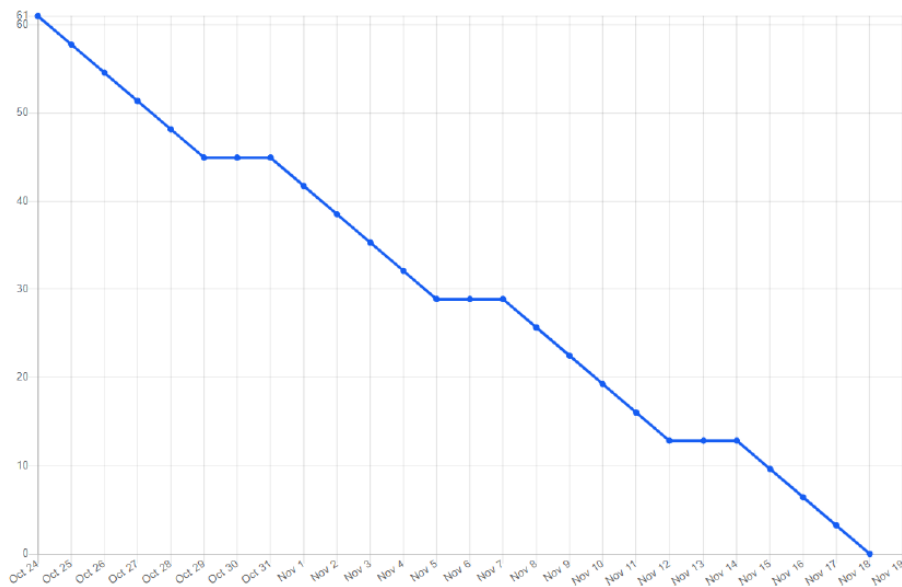
$$AV = \frac{\text{sprint duration}}{\text{velocity}} = \frac{20}{10} = 2$$

6.3 Reports from JIRA

Burndown Chart:

A burn down 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.


Burndown Chart




CODING AND SOLUTIONING

7.1 Feature 1

Firstly developed a jupyter notebook file then deployed in IBM Cloud by using services like watson studio, machine learning and cloud object












































 jupyter

Univerity admit eligibility predictor (2) Last Checkpoint: Last Wednesday at 10:50 AM (autosaved)

 Logout

FileEditViewInsertCellKernelWidgetsHelp

Connecting to kernelNot TrustedPython 3 (ipykernel)



PROJECT TITLE : UNIVERSITY ADMIT ELIGIBILITY PREDICTOR

TEAM ID : PNT2022TMID20129

TEAM MEMBERS: Dhanarajan G

Gokul Karthick P

Sanjay Kumar S

Ajay Rajendran S

IMPORT STATEMENTS

In [16]:

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

Load the data set

In [18]:

```
import os, types
import pandas as pd
from botocore.client import Config
import 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='7b1tUYzAq1AHKVEperhd9w-HV_fUQ34dEb3gPKu0jLpi',
                              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 = 'universityadmiteligibilitypredict-donotdelete-pr-jw2tqhsy1cmutv'
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 object
if not hasattr(body, "__iter__"): body.__iter__ = types.MethodType( __iter__, body )

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

Out[18]:

	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 [19]: data.drop(["Serial No."], axis=1, inplace=True)
```

```
In [20]: data.describe()
```

```
Out[20]:
```

	GRE Score	TOEFL Score	University Rating	SOP	LOR	CGPA	Research	Chance of Admit
count	400.000000	400.000000	400.000000	400.000000	400.000000	400.000000	400.000000	400.000000
mean	316.807500	107.410000	3.087500	3.400000	3.452500	8.598925	0.547500	0.724350
std	11.473646	6.069514	1.143728	1.006869	0.898478	0.596317	0.498362	0.142609
min	290.000000	92.000000	1.000000	1.000000	1.000000	6.800000	0.000000	0.340000
25%	308.000000	103.000000	2.000000	2.500000	3.000000	8.170000	0.000000	0.640000
50%	317.000000	107.000000	3.000000	3.500000	3.500000	8.610000	1.000000	0.730000
75%	325.000000	112.000000	4.000000	4.000000	4.000000	9.062500	1.000000	0.830000
max	340.000000	120.000000	5.000000	5.000000	5.000000	9.920000	1.000000	0.970000

```
In [21]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 400 entries, 0 to 399
Data columns (total 8 columns):
#   Column                Non-Null Count  Dtype
---  -
0   GRE Score              400 non-null   int64
1   TOEFL Score            400 non-null   int64
2   University Rating      400 non-null   int64
3   SOP                    400 non-null   float64
4   LOR                    400 non-null   float64
5   CGPA                   400 non-null   float64
6   Research               400 non-null   int64
7   Chance of Admit        400 non-null   float64
dtypes: float64(4), int64(4)
memory usage: 25.1 KB
```

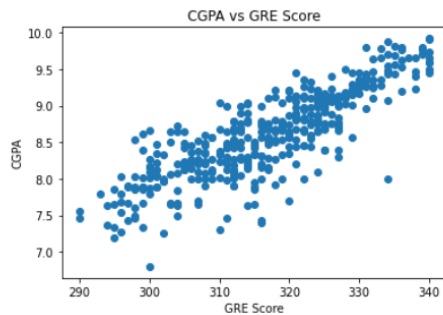
Handling Missing Values

```
In [22]: data.isnull().sum()
```

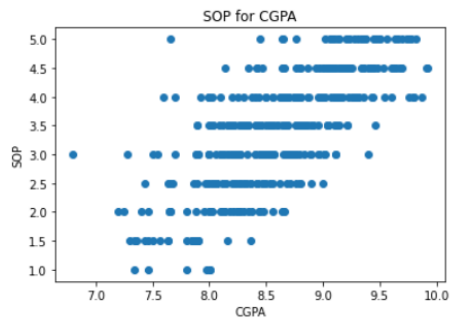
```
Out[22]: GRE Score      0
TOEFL Score      0
University Rating 0
SOP              0
LOR              0
CGPA             0
Research         0
Chance of Admit  0
dtype: int64
```

Visualization

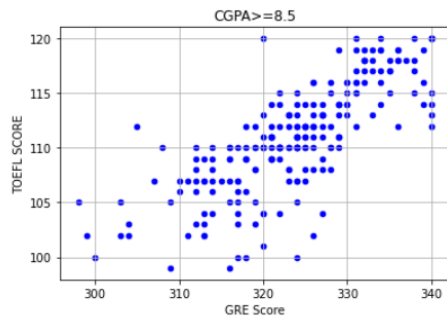
```
In [23]: plt.scatter(data['GRE Score'], data['CGPA'])
plt.title('CGPA vs GRE Score')
plt.xlabel('GRE Score')
plt.ylabel('CGPA')
plt.show()
```



```
In [24]: plt.scatter(data['CGPA'],data['SOP'])
plt.title('SOP for CGPA')
plt.xlabel('CGPA')
plt.ylabel('SOP')
plt.show()
```

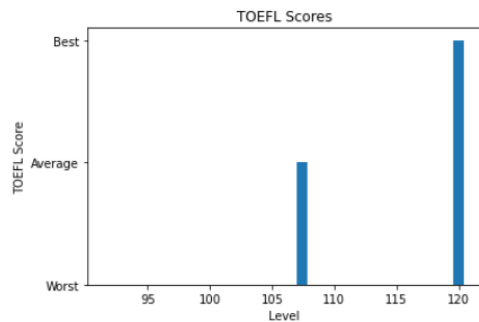


```
In [25]: data[data.CGPA >= 8.5].plot(kind='scatter', x='GRE Score', y='TOEFL Score',color="BLUE")
plt.xlabel("GRE Score")
plt.ylabel("TOEFL SCORE")
plt.title("CGPA>=8.5")
plt.grid(True)
plt.show()
```



```
In [26]: p = np.array([data["TOEFL Score"].min(),data["TOEFL Score"].mean(),data["TOEFL Score"].max()])
r = ["Worst","Average","Best"]
plt.bar(p,r)

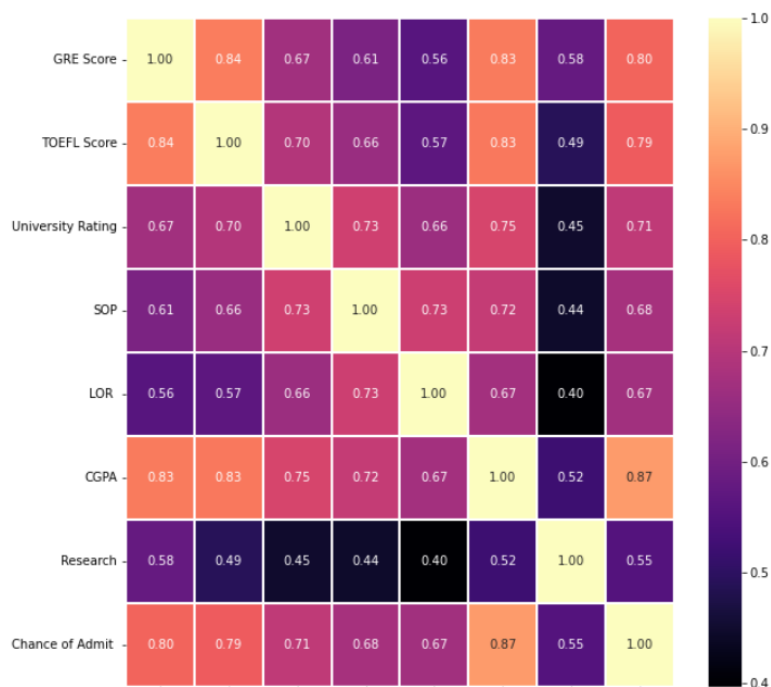
plt.title("TOEFL Scores")
plt.xlabel("Level")
plt.ylabel("TOEFL Score")
plt.show()
```



```
In [12]: plt.figure(figsize=(10, 10))

sns.heatmap(data.corr(), annot=True, linewidths=0.05, fmt= '.2f',cmap="magma")

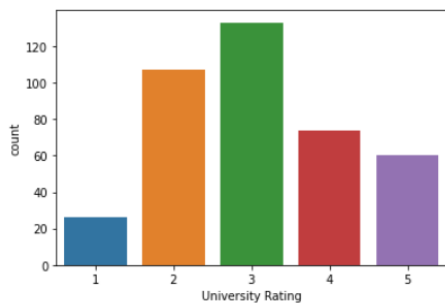
plt.show()
```



```
In [28]: data.Research.value_counts()

sns.countplot(x="University Rating",data=data)
```

```
Out[28]: <AxesSubplot:xlabel='University Rating', ylabel='count'>
```



Training and test split

```
In [29]: X=data.drop(['Chance of Admit '],axis=1) #input data_set
y=data['Chance of Admit '] #output Labels
```

```
In [30]: 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)
```


MODELING AND TRAINING

```
In [31]: from sklearn.ensemble import GradientBoostingRegressor
rgr = GradientBoostingRegressor()
rgr.fit(X_train,y_train)
```

```
Out[31]: GradientBoostingRegressor()
```

```
In [32]: rgr.score(X_test,y_test)
```

```
Out[32]: 0.7345575572947072
```

```
In [33]: y_predict=rgr.predict(X_test)
```

```
In [34]: 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)))
```

```
Mean Absolute Error: 0.048898416300819064
Mean Squared Error: 0.00443403167036226
Root Mean Squared Error: 0.06658852506522622
```

```
In [35]: y_train = (y_train>0.5)
y_test = (y_test>0.5)
```

```
In [36]: from sklearn.linear_model._logistic import LogisticRegression

lore = LogisticRegression(random_state=0, max_iter=1000)

lr = lore.fit(X_train, y_train)
```

```
In [38]: 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: 0.9333333333333333
Recall Score: 1.0
ROC AUC Score: 0.6666666666666667
Confussion Matrix:
[[ 2  4]
 [ 0 54]]
```

SAVING THE MODEL

```
In [39]: import pickle
```

```
In [40]: pickle.dump(lr, open("university.pkl", "wb")) #Logistic regression model
```

IBM Deployment

In [1]: `pip install -U ibm-watson-machine-learning`

```
Requirement already satisfied: ibm-watson-machine-learning in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (1.0.257)
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: 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: certifi in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (2022.9.24)
Requirement already satisfied: requests in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (2.26.0)
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: importlib-metadata in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (4.8.2)
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: tabulate in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (0.8.9)
Requirement already satisfied: lomond in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (0.3.3)
Requirement already satisfied: ibm-cos-sdk-core==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/python3.9/site-packages (from ibm-cos-sdk==2.11.*->ibm-watson-machine-learning) (0.10.0)
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: python-dateutil<3.0.0,>=2.1 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-cos-sdk-core==2.11.0->ibm-cos-sdk==2.11.*->ibm-watson-machine-learning) (2.8.2)
Requirement already satisfied: pytz>=2017.3 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (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/site-packages (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-packages (from python-dateutil<3.0.0,>=2.1->ibm-cos-sdk-core==2.11.0->ibm-cos-sdk==2.11.*->ibm-watson-machine-learning) (1.15.0)
Requirement already satisfied: charset-normalizer~2.0.0 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from requests->ibm-watson-machine-learning) (2.0.4)
Requirement already satisfied: idna<4,>=2.5 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from requests->ibm-watson-machine-learning) (3.3)
Requirement already satisfied: zipp>=0.5 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (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/python3.9/site-packages (from packaging->ibm-watson-machine-learning) (3.0.4)
Note: you may need to restart the kernel to use updated packages.
```

Authenticate and set space

In [2]: `from ibm_watson_machine_learning import APIClient
import json`

In [49]: `uml_credentials = {
 "url": "https://us-south.ml.cloud.ibm.com",
 "apikey": "PSONff0TD_cZHNECesojE_TSP73drzmK52IgxvaVTE1"
}

client = APIClient(uml_credentials)`

In [50]: `def guid_from_space_name(client, space_name):
 space = client.spaces.get_details()
 idr = []
 for i in space['resources']:
 idr.append(i['metadata']['id'])
 return idr`

In [68]: `space_uid = guid_from_space_name(client, "university")
print("Space Id:", space_uid[0])`

Space Id: dbdb029e-0578-4249-a7ef-a9afb6207be9

```
In [70]: client.software_specifications.list()
```

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
tensorflow_2.4-py3.7-horovod	1092590a-307d-563d-9b62-4eb7d64b3f22	base
pytorch_1.1-py3.6	10ac12d6-6b30-4ccd-8392-3e922c096a92	base
tensorflow_1.15-py3.6-dd1	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-474a5cfd5988	base
pytorch-onnx_rt22.1-py3.9-edt	1d362186-7ad5-5b59-8b6c-9d0880bde37f	base
tensorflow_2.1-py3.6	1eb25b84-d6ed-5dde-b6a5-3fbd1665666	base
spark-mllib_3.2	20047f72-0a98-58c7-9ff5-a77b012ebf53	base
tensorflow_2.4-py3.8-horovod	217c16f6-178f-56bf-824a-b19f20564c49	base
runtime-22.1-py3.9-cuda	26215f05-08c3-5a41-alb0-da66306ce658	base
do_py3.8	295addb5-9ef9-547e-9bf4-92ae3563e720	base
autoai-ts_3.8-py3.8	2aa0c932-798f-5ae9-abd6-15e0c2402fb5	base
tensorflow_1.15-py3.6	2b73a275-7cbf-420b-a912-eae7f436e0bc	base
kernel-spark3.3-py3.9	2b7961e2-e3b1-5a8c-a491-482c8368839a	base
pytorch_1.2-py3.6	2c8ef57d-2687-4b7d-acce-01f94976dac1	base
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
pytorch-onnx_1.2-py3.6-edt	40589d0e-7019-4e28-8daa-fb03b6f4fe12	base
pytorch-onnx_rt22.2-py3.10	40e73f55-783a-5535-b3fa-0c8b94291431	base
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

Save and Deploy the model

```
In [55]: import sklearn
sklearn.__version__
```

```
Out[55]: '1.0.2'
```

```
In [57]: MODEL_NAME = 'university'
DEPLOYMENT_NAME = 'uni'
DEMO_MODEL = 1r
```

```
In [61]: software_spec_uid = client.software_specifications.get_id_by_name('runtime-22.1-py3.9')
software_spec_uid
```

```
Out[61]: '12b83a17-24d8-5082-900f-0ab31fbfd3cb'
```

```
In [59]: model_props = {
    client.repository.ModelMetaNames.NAME: MODEL_NAME,
    client.repository.ModelMetaNames.TYPE: 'scikit-learn_1.0 ',
    client.repository.ModelMetaNames.SOFTWARE_SPEC_UID: software_spec_uid
}
```

```
In [65]: model_details = client.repository.store_model(
        model = DEMO_MODEL,
        meta_props = model_props,
        training_data = X_train,
        training_target = y_train
    )
    model_details
```

```
Out[65]: {'entity': {'hybrid_pipeline_software_specs': [],
'label_column': 'Chance of Admit ',
'schemas': {'input': [{'fields': [{'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': '2022-11-13T10:15:23.802Z',
'id': '62664227-d029-47f7-9981-abc041a29250',
'modified_at': '2022-11-13T10:15:26.309Z',
'name': 'university',
'owner': 'IBMid-6630020TFX',
'resource_key': '78014f14-04ce-48f0-b6a5-b749518a480a',
'space_id': 'dbdb029e-0578-4249-a7ef-a9afb6207be9'},
'system': {'warnings': []}}
```

```
In [71]: model_id = client.repository.get_model_id(model_details)
        model_id
```

```
Out[71]: '62664227-d029-47f7-9981-abc041a29250'
```

```
In [73]: deployment_props = {
        client.deployments.ConfigurationMetaNames.NAME: DEPLOYMENT_NAME,
        client.deployments.ConfigurationMetaNames.ONLINE: {}
    }

    deployment = client.deployments.create(
        artifact_uid = model_id,
        meta_props = deployment_props
    )
```

```
#####
```

```
Synchronous deployment creation for uid: '62664227-d029-47f7-9981-abc041a29250' 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='3610aaf5-cfd4-442e-b8e4-db7600a39f30'
-----
```

7.2 Feature 2

Developed an app.py file with integrated deployment and scoring points of IBM cloud.

```

from flask import Flask, render_template, redirect, url_for, request
import requests

app = Flask(__name__)

```

```

@app.route("/", methods=['POST', 'GET'])
def index():
    if request.method == 'POST':
        arr = []
        for i in request.form:
            val = request.form[i]
            if val == '':
                return redirect(url_for("demo2"))
            arr.append(float(val))

        # deepcode ignore HardcodedNonCryptoSecret: <please specify a reason of ignoring this>
        API_KEY = "Ps0NFf0TD_cZHNECesojE_TSP7JdrzmnKS2IgxvaVTE1"
        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}
        payload_scoring = {
            "input_data": [{"fields": ['GRE Score',
                                      'TOEFL Score',
                                      'University Rating',
                                      'SOP',
                                      'LOR ',
                                      'CGPA',
                                      'Research'],
                           "values": [arr]}]
        }
    }

```

```

        response_scoring = requests.post(
            'https://us-south.ml.cloud.ibm.com/ml/v4/deployments/83dcd36f-5b7d-42ba-aca0-8d88a3bc1f33/predictions?version=2022-11-42',
            json=payload_scoring,
            headers=header
        ).json()

        result = response_scoring['predictions'][0]['values']

        if result[0][0] > 0.5:
            return redirect(url_for('chance', percent=result[0][0] * 100))
        else:
            return redirect(url_for('no_chance', percent=result[0][0] * 100))
    else:
        return redirect(url_for("demo2"))

```

```

@app.route("/home")
def demo2():
    return render_template("demo2.html")

@app.route("/chance/<percent>")
def chance(percent):
    return render_template("chance.html", content=[percent])

@app.route("/noChance/<percent>")
def no_chance(percent):
    return render_template("noChance.html", content=[percent])

@app.route('/<path:path>')
def catch_all():
    return redirect(url_for("demo2"))

if __name__ == "__main__":
    app.run()

```

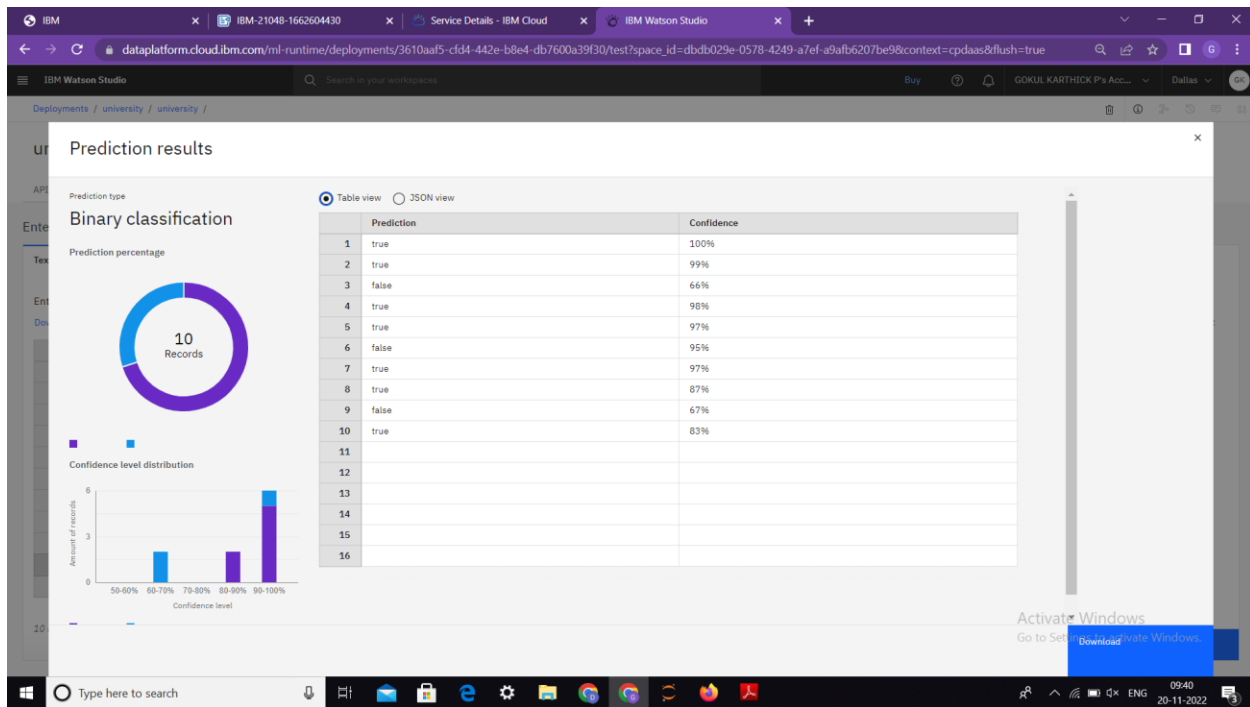
Testing

8.1 TestCases

The screenshot shows the IBM Watson Studio interface. At the top, there are tabs for IBM, IBM-21048-1662604430, Service Details - IBM Cloud, and IBM Watson Studio. The browser address bar shows a URL from dataplatform.cloud.ibm.com. The main interface displays the 'uni' model, which is 'Deployed' and 'Online'. Below this, there are tabs for 'API reference' and 'Test'. The 'Test' tab is active, showing an 'Enter input data' section. This section has two tabs: 'Text input' and 'JSON input'. The 'Text input' tab is selected, and it contains a table with 10 rows and 7 columns. The table headers are: GRE Score (int64), TOEFL Score (int64), University Rating (int64), SOP (float64), LOR (float64), CGPA (float64), and Research (int64). The table contains 10 rows of data, and the 11th row is highlighted with a blue border. At the bottom right of the table, there is a 'Predict' button. The bottom of the screen shows a Windows taskbar with various icons and the system clock showing 09:39 on 20-11-2022.

	GRE Score (int64)	TOEFL Score (int64)	University Rating (int64)	SOP (float64)	LOR (float64)	CGPA (float64)	Research (int64)
1	337	118	4	4.5	4.5	9.65	1
2	323	108	3	3.5	3	8.6	0
3	307	109	3	4	3	6	1
4	312	120	2	9	4	7	0
5	315	125	1	6	2.5	6	1
6	255	98	1	3	4	6.6	0
7	315	100	1	2	3.8	7.8	1
8	300	111	1	3.4	3.4	6.5	1
9	315	97	5	2	4.3	7	0
10	300	103	4	4.1	4.1	8	1
11							

If the student is eligible for the university it will give output as True. Otherwise the output will be False.



8.2 User Acceptance Testing

Purpose of Documentation :

The purpose of this documentation proved information, to give instructions, to persuade the reader, and to enact something.

Defect Analysis

Pre solution	Severty1	Severty2	Severty3	Severty4	Severty5
By Design	10	4	2	2	20
Duplicates	1	0	3	0	4
External	2	3	0	1	6
Fixed	11	2	4	20	37
Not reported	0	0	1	0	1
Skipped	0	0	1	1	2
Wont FLs	0	5	2	1	7
total	24	14	13	26	77

TestCase Analysis

section	testcases	Not tables	fall	pass
Print engine	7	0	0	7
Client application	31	0	0	81
security	2	0	0	2
Customer shipping	3	0	0	3
Exception Handling	9	0	0	9
Final report output	4	0	0	4
Version control	2	0	0	2

RESULTS

9.1 Performance Metrics

There are various metrics which we can use to evaluate the performance of ML algorithms, classification as well as regression algorithms. We must carefully choose the metrics for evaluating ML performance because –

- How the performance of ML algorithms is measured and compared will be dependent entirely on the metric you choose.
- How you weight the importance of various characteristics in the result will be influenced completely by the metric you choose.

```
In [47]: from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
from sklearn.metrics import classification_report
from sklearn.metrics import roc_auc_score
from sklearn.metrics import log_loss
results = confusion_matrix(y_test, y_pred)
print ('Confusion Matrix :')
print(results)
print ('Accuracy Score is',accuracy_score(y_test,y_pred))
print ('Classification Report : ')
print (classification_report(y_test, y_pred))
print('AUC-ROC:',roc_auc_score(y_test, y_pred))
print('LOGLOSS Value is',log_loss(y_test, y_pred))
```

```
Confusion Matrix :
[[ 4  2]
 [ 0 54]]
Accuracy Score is 0.9666666666666667
Classification Report :
              precision    recall  f1-score   support

   False       1.00      0.67      0.80         6
    True       0.96      1.00      0.98        54

 accuracy          0.97         60
  macro avg       0.98      0.83      0.89         60
 weighted avg     0.97      0.97      0.96         60

AUC-ROC: 0.8333333333333334
LOGLOSS Value is 1.1513191997446968
```

ADVANTAGES AND DISADVANTAGES

10.1 Advantages

- It helps students 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 Disadvantages

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

CONCLUSION

The subject of this examination was to determine if the below variables contribute to the admission of student to Master's degree program.

- GRE Score
- TOEFL
- Score
- University
- Rating
- SOP
- LOR
- CGPA

The results of this examination appear to indicate that it greatly contributes to the response variable 'Chance of Admit'. Higher the GRE, TOEFL score then higher the admit chances. The model predicts 91.5% accuracy and can be used for predicting the admit chances based on the above factors. This model will be helpful for the universities to predict the admission and ease their process of selection and timelines. As part of the hypothesis, the model proved that admission to Master's degree program is dependent on GRE, TOEFL and other scores. This model would likely be greatly improved by the gathering of additional data of students from different universities which has similar selection criteria to choose the candidates for Master's program.

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

Source Code GitHub & Project Demo Link

Source Code for Flask Application

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

app = Flask(__name__)

@app.route("/", methods=['POST', 'GET'])
def index():
    if request.method == 'POST':
        arr = []
        for i in request.form:
            val = request.form[i]
            if val == "":
                return redirect(url_for("demo2"))
            arr.append(float(val))

    # deepcode ignore HardcodedNonCryptoSecret: <please specify a reason of ignoring this>
    API_KEY = "PsONFf0TD_cZHNECesojE_TSP7JdrzmnKS2IgxvaVTE1"
    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}
    payload_scoring = {
        "input_data": [{"fields": ['GRE Score',
                                   'TOEFL Score',
                                   'University Rating',
                                   'SOP',
                                   'LOR ',
                                   'CGPA',
                                   'Research'],
                        "values": [arr]}]
    }

    response_scoring = requests.post(
        'https://us-south.ml.cloud.ibm.com/ml/v4/deployments/83dcd36f-5b7d-42ba-aca0-8d88a3be1f33/predictions?version=2022-11-13',
        json=payload_scoring,
        headers=header
    ).json()

    result = response_scoring['predictions'][0]['values']

    if result[0][0] > 0.5:
        return redirect(url_for('chance', percent=result[0][0] * 100))
    else:
        return redirect(url_for('no_chance', percent=result[0][0] * 100))
    else:
        return redirect(url_for("demo2"))

@app.route("/home")
def demo2():
```

```

return render_template("demo2.html")

@app.route("/chance/<percent>")
def chance(percent):
    return render_template("chance.html", content=[percent])

@app.route("/noChance/<percent>")
def no_chance(percent):
    return render_template("noChance.html", content=[percent])

@app.route("/<path:path>")
def catch_all():
    return redirect(url_for("demo2"))

if __name__ == "__main__":
    app.run()

```

Front End Code HTML Files

1.Index.html

```

{% extends 'index.html' %}

{% block body %}

<div class="container text-center p-4">
    <div class="d-flex justify-content-center">

        <div class="card" style="width: 34rem;">
            
            <div class="card-body">
                <h5 class="card-title">Congratulations!! You Have Chance to get admission</h5>
                <a href="/home" class="btn btn-primary">Go Back</a>
            </div>
        </div>
    </div>
</div>

{% endblock %}

```

2.Demo2.html

```

{% extends 'index.html' %}
{% block body %}
    <div class="p-4">
        <div class="row mb-3">
            <div class="col-4">
                <h1 class="text-responsive-h">
                    Enter your Score details and get the chance of your Admission
                </h1>
                <p class="text-responsive">

```

This Admission Prediction System will help the Students to check their Chance of Admission in Universities for their respective marks as they got in their final exams

```
</p>
<div class="d-flex justify-content-right">
  
</div>
</div>
<div class="col-8">
  <div class="card p-2 ms-2 my-2">
    <div class="card-body">
      <h5 class="card-title pb-4">
        <p style="text-align:center">Enter the Mark details</p>
      </h5>
      <form action="/" method="post" id="theForm">
        <div class="row mb-3">
          <label for="gre" class="col-lg-2 col-form-label">GRE Score:</label>
          <div class="col-lg-10">
            <input type="number" class="form-control" id="gre" name="gre" min="250" max="340" required>
          </div>
        </div>
        <div class="row mb-3">
          <label for="tofel" class="col-lg-2 col-form-label">TOFEL Score:</label>
          <div class="col-lg-10">
            <input type="number" class="form-control" id="tofel" name="tofel" min="50" max="120"
required>
          </div>
        </div>
        <div class="row mb-3">
          <label for="university_rating" class="col-lg-2 col-form-label">University Rating:</label>
          <div class="col-lg-10">
            <input type="number" class="form-control" id="university_rating" step="0.01"
name="university_rating" min="1" max="5" required>
          </div>
        </div>
        <div class="row mb-3">
          <label for="sop" class="col-lg-2 col-form-label">SOP:</label>
          <div class="col-lg-10">
            <input type="number" class="form-control" id="sop" name="sop" step="0.01" min="1" max="5"
required>
          </div>
        </div>
        <div class="row mb-3">
          <label for="lor" class="col-lg-2 col-form-label">LOR:</label>
          <div class="col-lg-10">
            <input type="number" class="form-control" id="lor" name="lor" step="0.01" min="1" max="5"
required>
          </div>
        </div>
        <div class="row mb-3">
          <label for="cgpa" class="col-lg-2 col-form-label">CGPA:</label>
          <div class="col-lg-10">
            <input type="number" class="form-control" id="cgpa" name="cgpa" step="0.01" min="5"
max="10" required>
          </div>
        </div>
        <fieldset class="row mb-3">
          <legend class="col-form-label col-sm-2 pt-0">Research:</legend>
          <div class="col-sm-10">
            <div class="form-check">
```

```

        <input class="form-check-input" type="radio" name="yes_no_radio" id="gridRadios1"
value="1">
        <label class="form-check-label" for="yes_no_radio">
        Yes
        </label>
    </div>
    <div class="form-check">
        <input class="form-check-input" type="radio" name="yes_no_radio" id="gridRadios2"
value="0" checked>
        <label class="form-check-label" for="yes_no_radio">
        No
        </label>
    </div>
</div>
</fieldset>
<div class="row lg-3">
    <div class="col-lg-2 mb-2 me-3">
        <button type="submit" class="btn btn-primary" id="button">Predict</button>
    </div>
    <div class="col-lg-2" id="spinner">
        <div class="spinner-border text-primary m-1" role="status">
            <span class="visually-hidden">Loading...</span>
        </div>
        <div class="spinner-grow text-primary m-1" role="status">
            <span class="visually-hidden">Loading...</span>
        </div>
    </div>
</div>
</form>
</div>
</div>
</div>
</div>
</div>
{ % endblock % }

<!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">
    <meta name="viewport" content="width=device-width, initial-scale=1, maximum-scale=1, user-scalable=no">
    <link rel="stylesheet" type="text/css" rel="noopener" target="_blank" href="../static/css/styles.css">
    <link href="https://cdn.jsdelivr.net/npm/bootstrap@5.2.2/dist/css/bootstrap.min.css" rel="stylesheet" integrity="sha384-
Zenh87qX5JnK2Jl0vWa8Ck2rdkQ2Bzep5IDxbcnCeuOxjzrPF/et3URy9Bv1WTRi" crossorigin="anonymous">
    <script type="text/javascript" src="../static/js/script.js" async></script>
    <title>University Admit Eligibility Predictor</title>
</head>
<body>
    <nav class="navbar navbar-expand-lg bg-light" >
        <div class="container-fluid">
            <a class="navbar-brand text-responsive-h" href="/">
                
                <marquee>University Admission Eligibility Prediction System</marquee>
            </a>
        </div>
    </nav>
    { % block body % }
    <h1> Index Page </h1>

```

```
{ % endblock % }
<script src="https://cdn.jsdelivr.net/npm/bootstrap@5.2.2/dist/js/bootstrap.bundle.min.js" integrity="sha384-OERcA2EqjJCMA+/3y+gxIOqMEjwtxJY7qPCqsdltbNJuaOe923+mo//f6V8Qbsw3" crossorigin="anonymous"></script>
</body>
</html>
```

3.chance.html

```
{% extends 'index.html' %}
{% block body %}

<div class="container text-center p-4">
  <div class="d-flex justify-content-center">
    <div class="card" style="width: 34rem;">
      
      <div class="card-body">
        <h5 class="card-title">OOPS!! You have a LOW / NO chance to get admission </h5>
        <p class="card-text">The model has predicted that you have no chance</p>
        <a href="/home" class="btn btn-primary">Go Back</a>
      </div>
    </div>
  </div>
</div>

{% endblock %}
```

4.nochance.html

```
{% extends 'index.html' %}
{% block body %}

<div class="container text-center p-4">
  <div class="d-flex justify-content-center">
    <div class="card" style="width: 34rem;">
      
      <div class="card-body">
        <h5 class="card-title">OOPS!! You have a LOW / NO chance to get admission </h5>
        <p class="card-text">The model has predicted that you have no chance</p>
        <a href="/home" class="btn btn-primary">Go Back</a>
      </div>
    </div>
  </div>
</div>

{% endblock %}
```

5.Styles.css

```
* {
  margin: 0;
  padding: 0;
  border: 0;
}
body {
```

```

font: 62.5%/1.5 "Lucida Grande", "Lucida Sans", Tahoma, Verdana, sans-serif;
background: #e0eafc;
background: -webkit-linear-gradient(to right, #e0eafc, #cfdef3);
background: linear-gradient(to right, #e0eafc, #cfdef3);
color: #000000;
text-align:center;
}

h1 {
  font-size: 2.2em;
}

h2 {
  font-size: 2.0em;
}

h4 {
  font-size: 1.6em;
}

p {
  font-size: 1.2em;
}

input.text
{
  padding: 3px;
  border: 1px solid #999999;
}
.p-4{
background-color:#ffa07a;
}
img {
  max-width: auto;
  height: auto;
}

.text-responsive {
  font-size: calc(50% + 0.6vw + 0.6vh);
}

.text-responsive-h {
  font-size: calc(80% + 0.6vw + 0.6vh);
}
Footer

```

6.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 = "Checking..."
loading.style.display = "block"
};


const enableButton = () => {
  console.log("Loading window...");
  button.disabled = false;
  button.className = "btn btn-primary"
  button.innerHTML = "Check your chance"
  loading.style.display = "none"
}

theForm.onsubmit = disableButton;


window.onload = enableButton;

```

6 Jupyter Notebook


jupyter












Univerity admit eligibility predictor (2) Last Checkpoint: Last Wednesday at 10:50 AM (autosaved)


Logout

File Edit View Insert Cell Kernel Widgets Help

Connecting to kernel Not Trusted

Python 3 (ipykernel)












 Markdown

PROJECT TITLE : UNIVERSITY ADMIT ELIGIBILITY PREDICTOR

 TEAM ID : PNT2022TMID20129

 TEAM MEMBERS: Dhanarajan G

 Gokul Karthick P

 Sanjay Kumar S

 Ajay Rajendran S

 IMPORT STATEMENTS

In [16]: `import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline`

Load the data set

```
In [18]: import os, types
import pandas as pd
from botocore.client import Config
import 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='7bitUYzAq1AHKVEperhd9w-HV_fUQ34dEb3gPKu0jLpi',
                              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 = 'universityadmiteligibilitypredict-donotdelete-pr-jw2tqhsy1cmutv'
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 object
if not hasattr(body, "__iter__"): body.__iter__ = types.MethodType(__iter__, body)

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

```
Out[18]:
```

	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 [19]: data.drop(["Serial No."], axis=1, inplace=True)
```

```
In [20]: data.describe()
```

```
Out[20]:
```

	GRE Score	TOEFL Score	University Rating	SOP	LOR	CGPA	Research	Chance of Admit
count	400.000000	400.000000	400.000000	400.000000	400.000000	400.000000	400.000000	400.000000
mean	316.807500	107.410000	3.087500	3.400000	3.452500	8.598925	0.547500	0.724350
std	11.473646	6.069514	1.143728	1.006869	0.898478	0.596317	0.498362	0.142609
min	290.000000	92.000000	1.000000	1.000000	1.000000	6.800000	0.000000	0.340000
25%	308.000000	103.000000	2.000000	2.500000	3.000000	8.170000	0.000000	0.640000
50%	317.000000	107.000000	3.000000	3.500000	3.500000	8.610000	1.000000	0.730000
75%	325.000000	112.000000	4.000000	4.000000	4.000000	9.062500	1.000000	0.830000
max	340.000000	120.000000	5.000000	5.000000	5.000000	9.920000	1.000000	0.970000

```
In [21]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 400 entries, 0 to 399
Data columns (total 8 columns):
#   Column                Non-Null Count  Dtype
---  -
0   GRE Score              400 non-null   int64
1   TOEFL Score            400 non-null   int64
2   University Rating      400 non-null   int64
3   SOP                    400 non-null   float64
4   LOR                    400 non-null   float64
5   CGPA                   400 non-null   float64
6   Research               400 non-null   int64
7   Chance of Admit        400 non-null   float64
dtypes: float64(4), int64(4)
memory usage: 25.1 KB
```

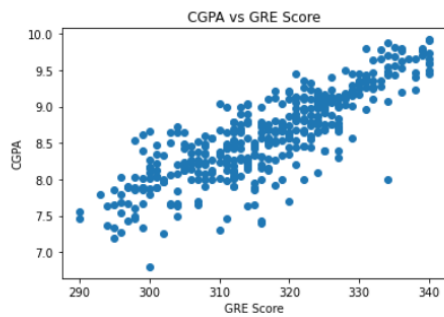
Handling Missing Values

```
In [22]: data.isnull().sum()
```

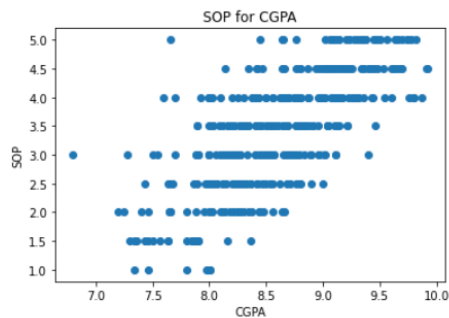
```
Out[22]: GRE Score      0
         TOEFL Score    0
         University Rating 0
         SOP            0
         LOR           0
         CGPA          0
         Research       0
         Chance of Admit 0
         dtype: int64
```

Visualization

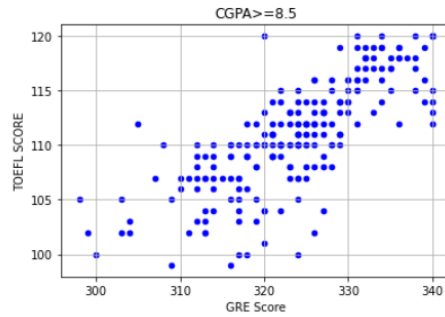
```
In [23]: plt.scatter(data['GRE Score'],data['CGPA'])
         plt.title('CGPA vs GRE Score')
         plt.xlabel('GRE Score')
         plt.ylabel('CGPA')
         plt.show()
```



```
In [24]: plt.scatter(data['CGPA'],data['SOP'])
         plt.title('SOP for CGPA')
         plt.xlabel('CGPA')
         plt.ylabel('SOP')
         plt.show()
```

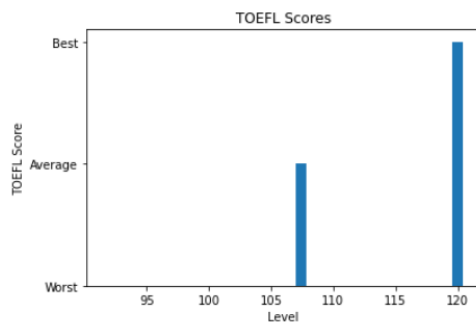


```
In [25]: data[data.CGPA >= 8.5].plot(kind='scatter', x='GRE Score', y='TOEFL Score',color="BLUE")
plt.xlabel("GRE Score")
plt.ylabel("TOEFL SCORE")
plt.title("CGPA>=8.5")
plt.grid(True)
plt.show()
```

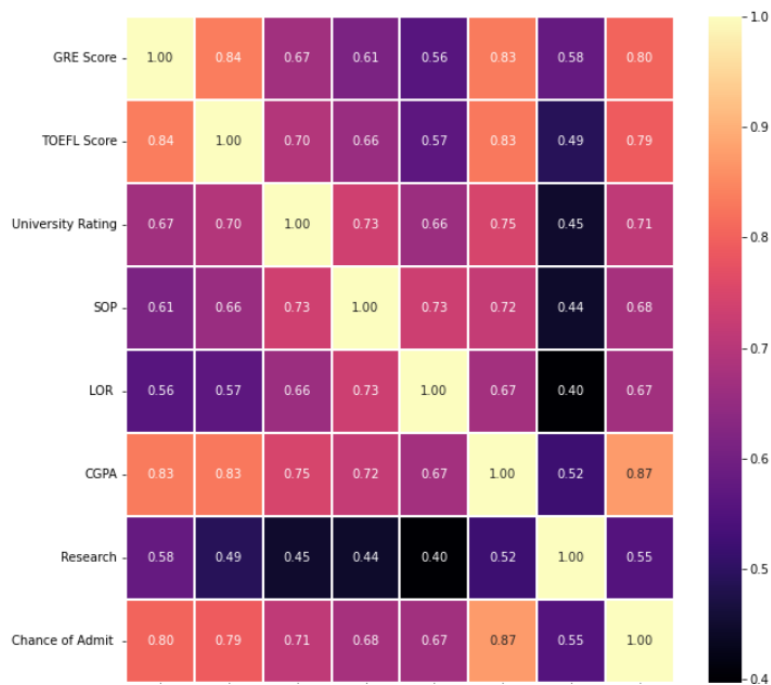


```
In [26]: p = np.array([data["TOEFL Score"].min(),data["TOEFL Score"].mean(),data["TOEFL Score"].max()])
r = ["Worst","Average","Best"]
plt.bar(p,r)

plt.title("TOEFL Scores")
plt.xlabel("Level")
plt.ylabel("TOEFL Score")
plt.show()
```

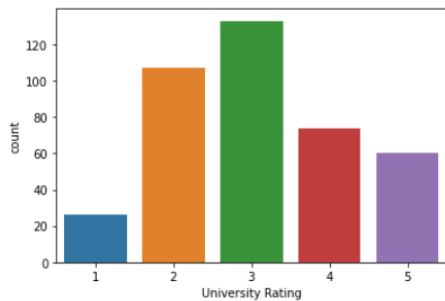


```
In [12]: plt.figure(figsize=(10, 10))
sns.heatmap(data.corr(), annot=True, linewidths=0.05, fmt='%.2f', cmap="magma")
plt.show()
```



```
In [28]: data.Research.value_counts()
sns.countplot(x="University Rating", data=data)
```

```
Out[28]: <AxesSubplot: xlabel='University Rating', ylabel='count'>
```



Training and test split

```
In [29]: X=data.drop(['Chance of Admit '],axis=1) #input data_set
y=data['Chance of Admit '] #output Labels
```

```
In [30]: 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)
```

MODELING AND TRAINING

```
In [31]: from sklearn.ensemble import GradientBoostingRegressor
rgr = GradientBoostingRegressor()
rgr.fit(X_train,y_train)
```

```
Out[31]: GradientBoostingRegressor()
```

```
In [32]: rgr.score(X_test,y_test)
```

```
Out[32]: 0.7345575572947072
```

```
In [33]: y_predict=rgr.predict(X_test)
```

```
In [34]: 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)))
```

```
Mean Absolute Error: 0.048898416300819064
Mean Squared Error: 0.00443403167036226
Root Mean Squared Error: 0.06658852506522622
```

```
In [35]: y_train = (y_train>0.5)
y_test = (y_test>0.5)
```

```
In [36]: from sklearn.linear_model._logistic import LogisticRegression
```

```
lore = LogisticRegression(random_state=0, max_iter=1000)
lr = lore.fit(X_train, y_train)
```

```
In [38]: 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: 0.9333333333333333
Recall Score: 1.0
ROC AUC Score: 0.6666666666666667
Confussion Matrix:
[[ 2  4]
 [ 0 54]]
```

SAVING THE MODEL

```
In [39]: import pickle
```

```
In [40]: pickle.dump(lr, open("university.pkl", "wb")) #Logistic regression model
```

IBM Deployment

```
In [1]: pip install -U ibm-watson-machine-learning
```

```
Requirement already satisfied: ibm-watson-machine-learning in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (1.0.257)
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: 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: certifi in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (2022.9.24)
Requirement already satisfied: requests in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (2.26.0)
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: importlib-metadata in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (4.8.2)
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: tabulate in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (0.8.9)
Requirement already satisfied: lomond in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (0.3.3)
Requirement already satisfied: ibm-cos-sdk-core==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/python3.9/site-packages (from ibm-cos-sdk==2.11.*->ibm-watson-machine-learning) (0.10.0)
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: python-dateutil<3.0.0,>=2.1 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-cos-sdk-core==2.11.0->ibm-cos-sdk==2.11.*->ibm-watson-machine-learning) (2.8.2)
Requirement already satisfied: pytz>=2017.3 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (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/site-packages (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-packages (from python-dateutil<3.0.0,>=2.1->ibm-cos-sdk-core==2.11.0->ibm-cos-sdk==2.11.*->ibm-watson-machine-learning) (1.15.0)
Requirement already satisfied: charset-normalizer~=2.0.0 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from requests->ibm-watson-machine-learning) (2.0.4)
Requirement already satisfied: idna<4,>=2.5 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from requests->ibm-watson-machine-learning) (3.3)
Requirement already satisfied: zipp>=0.5 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (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/python3.9/site-packages (from packaging->ibm-watson-machine-learning) (3.0.4)
Note: you may need to restart the kernel to use updated packages.
```

Authenticate and set space

```
In [2]: from ibm_watson_machine_learning import APIClient
import json
```

```
In [49]: uml_credentials = {
    "url": "https://us-south.ml.cloud.ibm.com",
    "apikey": "PSONff0TD_cZHNECesojE_TSP73drzmnKS2IgxvaVTE1"
}

client = APIClient(uml_credentials)
```

```
In [50]: def guid_from_space_name(client, space_name):
    space = client.spaces.get_details()
    idr = []
    for i in space['resources']:
        idr.append(i['metadata']['id'])
    return idr
```

```
In [68]: space_uid = guid_from_space_name(client, "university")
print("Space Id:", space_uid[0])
```

Space Id: dbdb029e-0578-4249-a7ef-a9afb6207be9

```
In [70]: client.software_specifications.list()
```

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	09f4c4f0-90a7-5899-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
tensorflow_2.4-py3.7-horovod	1092590a-307d-563d-9b62-4eb7d64b3f22	base
pytorch_1.1-py3.6	10ac12d6-6b30-4ccd-8392-3e922c096a92	base
tensorflow_1.15-py3.6-dd1	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-b6a5-3fbdff166566	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
autoai-ts_3.8-py3.8	2aa0c932-798f-5ae9-abd6-15e0c2402fb5	base
tensorflow_1.15-py3.6	2b73a275-7cbf-420b-a912-eae7f436e0bc	base
kernel-spark3.3-py3.9	2b7961e2-e3b1-5a8c-a491-482c8368839a	base
pytorch_1.2-py3.6	2c8ef57d-2687-4b7d-acce-01f94976dac1	base
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
pytorch-onnx_1.2-py3.6-edt	40589d0e-7019-4e28-8daa-fb03b6f4fe12	base
pytorch-onnx_rt22.2-py3.10	40e73f55-783a-5535-b3fa-0c8b94291431	base
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

Save and Deploy the model

```
In [55]: import sklearn  
sklearn.__version__
```

```
Out[55]: '1.0.2'
```

```
In [57]: MODEL_NAME = 'university'  
DEPLOYMENT_NAME = 'uni'  
DEMO_MODEL = lr
```

```
In [61]: software_spec_uid = client.software_specifications.get_id_by_name('runtime-22.1-py3.9')  
software_spec_uid
```

```
Out[61]: '12b83a17-24d8-5082-900f-0ab31fbfd3cb'
```

```
In [59]: model_props = {  
    client.repository.ModelMetaNames.NAME: MODEL_NAME,  
    client.repository.ModelMetaNames.TYPE: 'scikit-learn_1.0 ',  
    client.repository.ModelMetaNames.SOFTWARE_SPEC_UID: software_spec_uid  
}
```

```
In [71]: model_id = client.repository.get_model_id(model_details)  
model_id
```

```
Out[71]: '62664227-d029-47f7-9981-abc041a29250'
```

```
In [73]: deployment_props = {  
    client.deployments.ConfigurationMetaNames.NAME: DEPLOYMENT_NAME,  
    client.deployments.ConfigurationMetaNames.ONLINE: {}  
}  
  
deployment = client.deployments.create(  
    artifact_uid = model_id,  
    meta_props = deployment_props  
)
```

```
#####
```

Synchronous deployment creation for uid: '62664227-d029-47f7-9981-abc041a29250' 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='3610aaf5-cfd4-442e-b8e4-db7600a39f30'  
-----
```

Project Demo Link:

https://drive.google.com/file/d/10Nr5LMu5uA5kKIKTS8l4C-koUeE6heCt/view?usp=share_link

Github Link:

<https://github.com/IBM-EPBL/IBM-Project-21048-1659771001>