IBM Project Report

University Admit Eligibility Predictor

1. Introduction

1. Project Overview

Concerns about getting into college are common among students. This project's goal is to assist students in narrowing down institutions based on their profiles. They have a good idea based on the expected results regarding the likelihood of getting into a specific university. This analysis ought to provide better insight for students who are or will be preparing for exams and building their profiles. The University Admit Predictor is a web-based program that allows students to register with their personal information and academic record to predict college admission. The main motivation for this project is to help students who cannot afford highly-priced predictor systems. This project will reduce the costs of applications for aspirants significantly because an aspirant will be able to know the college, he has a chance at before actually applying to that college

2. Purpose

The overall purpose is to talk about how admittance to a university can be predicted using a variety of variables using linear regression. For postgraduates, many potential students submit applications. The admittance decision is based on requirements set forth by the specific college or degree program. In order to forecast graduate school acceptance, the independent factors in this study will be statistically measured. If successful, exploration and data analysis would enable predictive models to better prioritize the application screening process for Master's degree programs, resulting in the admission of the most qualified applicants.

2. Literature Survey

1. Existing Problem

The existing problem with the available predictors is that most of them charge way too much money for predicted results. They also as for too many details like where did you go schooling? What was your childhood interest? etc. These processes are time-consuming.

2. References

- [1] M. Omaer Faruq Goni, A. Matin, T. Hasan, M. Abu Ismail Siddique, O. Jyoti and F. M. Sifnatul Hasnain, "Graduate Admission Chance Prediction Using Deep Neural Network," 2020 IEEE International Women in Engineering (WIE) Conference on Electrical and Computer Engineering (WIECON-ECE), 2020.
- [2] Nandal, P., Deep Learning in diverse Computing and Network Applications Student Admission Predictor using Deep Learning (March 28, 2020). Proceedings of the International Conference on Innovative Computing & Communications (ICICC) 2020.
- [3] H. Fathiya and L. Sadath, "University Admissions Predictor Using Logistic Regression," 2021 International Conference on Computational Intelligence and Knowledge Economy (ICCIKE), 2021.

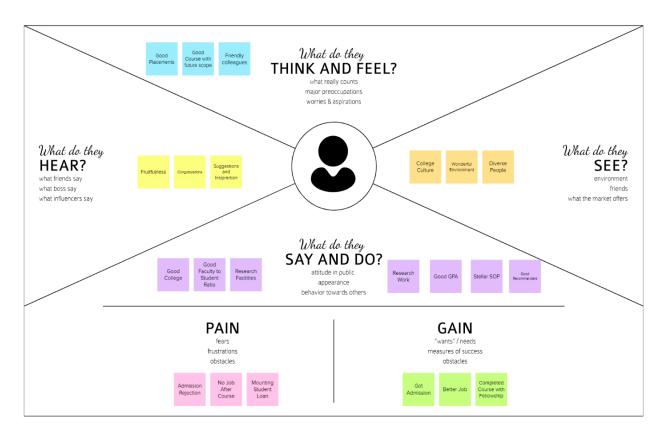
- [4] S. Sridhar, S. Mootha and S. Kolagati, "A University Admission Prediction System using Stacked Ensemble Learning," 2020 Advanced Computing and Communication Technologies for High Performance Applications (ACCTHPA), 2020.
- [5] Z. Bitar and A. Al-Mousa, "Prediction of Graduate Admission using Multiple Supervised Machine Learning Models," 2020 SoutheastCon, 2020.

2.3 Problem Statement Definition

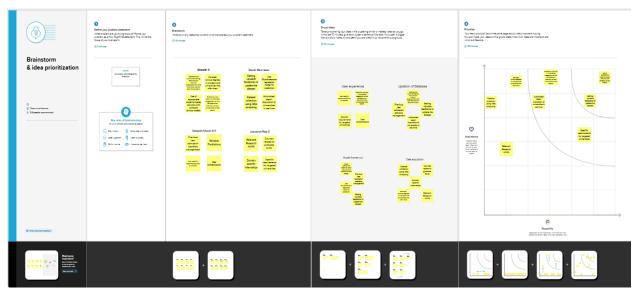
Creating a program that uses machine learning models to forecast a student's prospects of admission to a university. The model will be trained and stored on the IBM cloud. Prediction of student admission to university based on their profile.

3. Ideation and Proposed Solution

3.1 Empathy Map Canvas



3.2 Ideation & Brainstorming



3.3 Proposed Solution

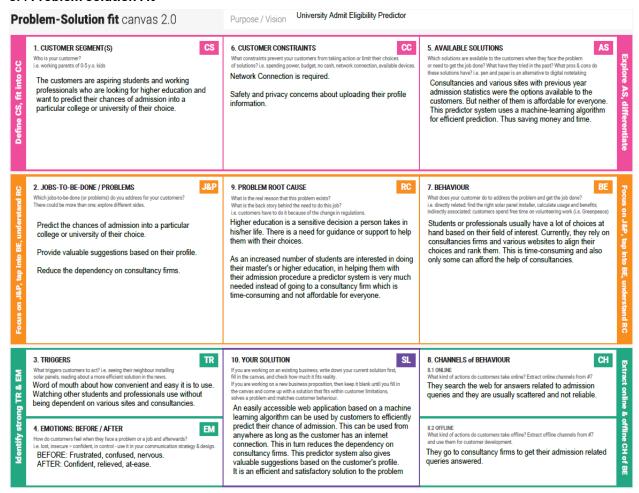
Proposed Solution Template

Date	27 September 2022
Team ID	PNT2022TMID52898
Project Name	University Admit Eligibility Predictor

S.No.	Parameter	Description
	Problem Statement (Problem	toPrediction of student admission
1.	be solved)	to university based on their
		profile.
	Idea / Solution description	In order to guide the students
2.		with their admission procedure a
		Machine Learning based
		predictor system has to be
		introduced to provide students
		with best possible predictions.

	Novelty / Uniqueness	So far there are very few
3.		solutions that exists which is
		similar to our proposed solutions.
		Yocket, College AI, GradCafe
		are a few.
		But all those have their own
		drawbacks in terms of usability
		and Output. Our solution tries to
		address these drawbacks in an
		efficient manner.
	Social Impact / Customer	A customer would be able to
4.	Satisfaction	predict the college he will get
		beforehand so he will not get
		disappointed if he didn't get his
		dream college. It will save
		money for customers since he
		will apply for colleges in which
		he has a better chance of
		admission.
	Business Model (Revenue	This predictor system can be
5.	Model)	used by consultancy firms to help
		their customers who seek their
		help.
		Apart from this valuable
		suggestion can be provided for
		an appropriate subscription fee.
	Scalability of the Solution	The solution could be scaled up
6.		to include more universities and
		larger geographical area. But
		maintenance of large database
		might pose a few problems.

3.4 Problem Solution Fit



4. Requirement Analysis

4.1 Functional Requirement

Following are the functional	Functional Requirement	Sub Requirement (Story /
requirements of the proposed	(Epic)	Sub-Task)
solution. FR No.		
FR-1	User Registration	To prevent unauthorised access,
		users must be able to login
		using their email Id and
		password.
FR-2	User Confirmation	Confirmation via Email
		Confirmation via OTP

FR-3	User Details	Submit the documents such as:		
		1. GRE, TOEFLT/IELTS		
		Scorecard,		
		2. Curriculum Vitae,		
		3. Letter of Recommendation		
		(LOR)		
		4. Statement of Purpose (SOP)		
FR-4	User Profile	User Dashboard containing		
		his/her personal information,		
		wish list, skills, and hobbies.		
FR-5	Data Management	A user can create, read, update,		
		and delete data.		

4.2 Non-Functional Requirement

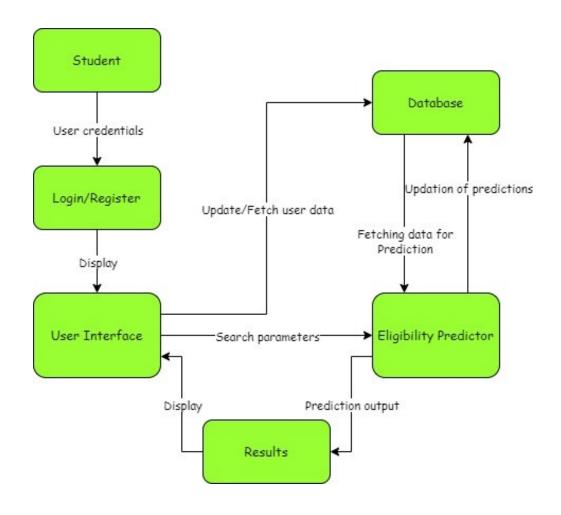
FR No.	Non-Functional Rec	quirement Description
NFR-1	Usability	 Proposed solution should be user friendly. The system should not require any prior knowledge from the user. The system should be able to load data quickly.
NFR-2	Security	 Only the authorized users can use the site's services. Some cryptographic techniques need to be used for validation purposes.

NFR-3	Reliability	 Data backups and strategies are to be used to avoid data being lost or data being corrupted. They system should be functional at any time of the day.
NFR-4	Performance	 At any instant the system should be able to support multiple users. The prediction made for user requests must not take more time. Preferably it should be less than 5 seconds.
NFR-5	Availability	The system should be functional at any time of the day but in case of error, it should display backup page and retrieve information from backup folder.
NFR-6	Scalability	Assesses the highest workloads under which the system will still meet the performance Deals with the measure of the system's response time under different load conditions requirements.

5. **Project Design**

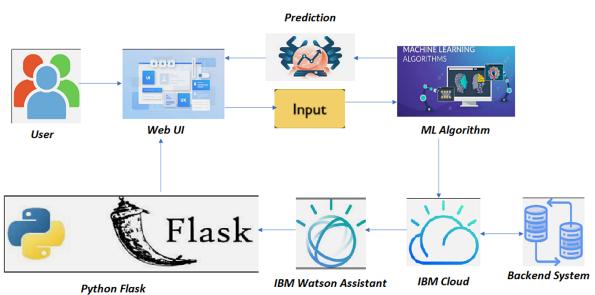
5.1 Data Flow Diagrams

The traditional visual representation of how information moves through a system is a data flow diagram (DFD). A tidy and understandable DFD can graphically represent the appropriate quantity of the system demand. It demonstrates how information enters and exits the system, what modifies the data, and where information is kept.

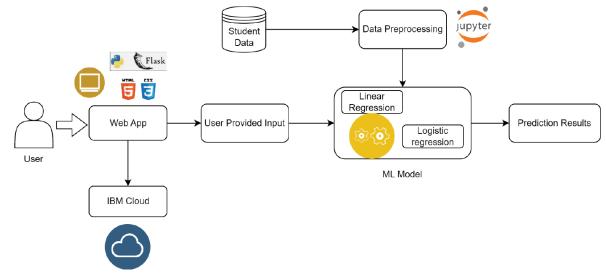


5.2 Solution & Technical Architecture

Solution Architecture



Technical Architecture



5.3 User Stories

User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance criteria	Priority	Release
Student	Registration	USN-1	User can register by entering the username, email ID and password	I can access my account / dashboard	Medium	Sprint-2
	Login USN-2 As a user, I can log into the application by entering email & password Update USN-3 Profile USN-3 Dystretion Loging in post registration, I will		I can access my account	Medium	Sprint-2	
		USN-3		I can complete the profile updation by filling in details	Medium	Sprint-2
	Prediction Results	USN-4	Exploratory Data Analysis and predict chance of admission using ML models.	I can see the prediction results of chances	High	Sprint-1
	Integrating application with ML model	USN-5	User can get chance of admission after entering the details	I can get chance of admit by filling in the details. The pickle object is integrated with the application	Medium	Sprint-3
	Submit admission data	USN-6	Can submit their admission results with their profile details	I can submit my admission data and receive a confirmation that my data is submitted	Medium	Sprint-4
	IBM Watson Deployment	USN-7	Integrating our application with IBM cloud and deploying it.	I can get reliable prediction in the webpage via the model deployed in cloud.	Medium	Sprint-4

6. Project Planning & Scheduling6.1 Sprint Planning & Estimation

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	Exploratory Data Analysis	USN-4	Visualizing the dataset and analysing the various trends.	2	High	Bharath R, Daniel Mark Isaac
Sprint-1	Model Building	USN-4	Developing a ML model to predict the chance of admission using the dataset	1	High	Bharath R, Daniel Mark Isaac
Sprint-2	Login	USN-2 USN-3	After login user can update their profile and start predicting their chance of admission	2	Medium	Bharath R, Deepakh Sharan D P
Sprint-2	Register	USN-1	After registering, user can login to view his/her account.	2	Medium	Bharath R, Deepakh Sharan D P
Sprint-3	Application Building	USN-1	Integrate the application with pickle object of the model	1	Low	Deepakh Sharan D P, Jaswandt Raja S
Sprint-3	Application Building	USN-5	User can get the chance of admission after entering the details	1	Low	Deepakh Sharan D P, Jaswandt Raja S
Sprint-4	IBM Watson Deployment	USN-7	Integrating our application with IBM cloud and deploying it.	2	Medium	Daniel Mark Isaac, Jaswandt Raja S

6.2 Sprint Delivery Schedule

Sprint	Total Story Points	Duration	Sprint Start Date	Sprint End Date (Planned)	Story Points Completed (as on Planned End Date)	Sprint Release Date (Actual)
Sprint-1	20	6 Days	24 Oct 2022	29 Oct 2022	20	15 Nov 22
Sprint-2	20	6 Days	31 Oct 2022	05 Nov 2022	20	16 Nov 22
Sprint-3	20	6 Days	07 Nov 2022	12 Nov 2022	20	17 Nov 22
Sprint-4	20	6 Days	14 Nov 2022	19 Nov 2022	20	18 Nov 22

7. Coding and Solutioning

7.1 Feature 1

Training The Linear Regression model:

A Linear regression model is trained on the dataset:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.linear_model import LinearRegression

dataset = pd.read_csv("Admission_Predict.csv")
dataset.head()

x = dataset[['GRE Score','TOEFL Score','University Rating','SOP','LOR ','CGPA','Research']]
y = dataset['Chance of Admit ']

from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.3, random_state = 100)

mlr = LinearRegression()
mlr.fit(x_train, y_train)

print("Intercept: ", mlr.intercept_)
print("Coefficients:")
list(zip(x, mlr.coef_))
```

Testing the model and evaluating performance metrics:

```
#Prediction of test set

y_pred_mlr= mlr.predict(x_test)

#Predicted values
print("Prediction for test set: {}".format(y_pred_mlr))

mlr_diff = pd.DataFrame({'Actual value': y_test, 'Predicted value': y_pred_mlr})
mlr_diff.head()

from sklearn import metrics
meanAbErr = metrics.mean_absolute_error(y_test, y_pred_mlr)
meanSqErr = metrics.mean_squared_error(y_test, y_pred_mlr)
rootMeanSqErr = np.sqrt(metrics.mean_squared_error(y_test, y_pred_mlr))
print('R squared: {:.2f}'.format(mlr.score(x,y)*100))
print('Mean Absolute Error:', meanAbErr)
print('Mean Square Error:', meanSqErr)
print('Root Mean Square Error:', rootMeanSqErr)
```

```
Intercept: -1.228285398131014
Coefficients:
Prediction for test set: [0.79793967 0.74792724 0.86871864 0.8856855 0.68403746 0.85370856
0.67161949 0.7408362 0.90663309 0.58312346 0.69567936 0.70838948
0.63860073 0.67760163 0.61539916 0.66737881 0.86000658 0.60810758
0.57801389 0.83481169 0.63781003 0.84700949 0.48541665 0.80094769
0.46753179 0.74748586 0.73368904 0.77987137 0.53655206 0.95974245
0.89124757 0.77355509 0.7263695 0.7001736 0.55882143 0.52577021
0.59819879 0.73301941 0.9236683 0.48339279 0.66735848 0.65187014
0.78597154 0.59528278 0.77790546 0.41384251 0.61783883 0.88046811
0.55779058 0.56629417 0.93355935 0.66785998 0.82618029 0.80313414
0.74052551 0.85051165 0.42295811 0.68660609 0.78066541 0.70283069
0.81316624 0.52125707 0.63282489 0.8123818 0.61293148 0.73399553
0.82449109 0.71593313 0.63620347 0.69537323 0.90186005 0.63823487
0.74906378 0.65595495 0.8590957 0.53615055 0.67875334 0.6856359
0.67688479 0.85390393 0.84227233 0.90884758 0.69405595 0.48302994
0.93644606 0.70666517 0.79116822 0.84675279 0.62712644 0.52163612
0.59865886 0.44918123 0.70366479 0.84887413 0.45395457 0.77122787
0.83264783 0.68634576 0.64283562 0.7096491 0.68226406 0.80400285
0.96883608 0.87650171 0.59008477 0.72890014 0.59882941 0.64737974
0.60326834 0.63152619 0.69819591 0.59452426 0.85136712 0.92622498
0.83075228 0.64479398 0.62520864 0.59349123 0.6992843 0.68393413]
R squared: 80.17
Mean Absolute Error: 0.04692878285488935
Mean Square Error: 0.003716513784197384
Root Mean Square Error: 0.060963216649036686
```

Deploying the model in cloud:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.linear_model import LinearRegression
from sklearn import model_selection
import warnings
warnings.filterwarnings("ignore")
```

```
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 y
# You might want to remove those credentials before you share the notebook.
cos_client = ibm_boto3.client(service_name='s3',
    ibm_api_key_id='o7_Ng7bC445bx9xZCqECGhEgfxkZvsBlT42_odbLaFmK',
    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 = 'eligibilitypredictionmodelling-donotdelete-pr-dywlho4m5azqwc'
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 )
dataset = pd.read_csv(body)
dataset.head()
```

	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

```
dataset.drop('Serial No.',axis = 'columns', inplace = True)
dataset.head()
```

	GRE Score	TOEFL Score	University Rating	g SC	OP L	.OR	CGPA	Research	Chance of A	dmit
0	337	118		4 4	1.5	4.5	9.65		1	0.92
1	324	107		4 4	1.0	4.5	8.87	-	1	0.76
2	316	104	:	3 3	3.0	3.5	8.00	•	1	0.72
3	322	110	:	3 3	3.5	2.5	8.67	-	1	0.80
	GRE Score	TOEFL Score	University Rating	SOP	LOR	CG	PA Res	search C	hance of Admit	
4	314	103	2	2.0	3.0	8 (.21	0	0.65	

```
x = dataset[['GRE Score','TOEFL Score','University Rating','SOP','LOR ','CGPA','Rese
y = dataset['Chance of Admit ']
```

```
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.3, random_st

mlr = LinearRegression()
mlr.fit(x_train, y_train)
```

LinearRegression()

```
y_pred_mlr= mlr.predict(x_test)

from sklearn import metrics
meanAbErr = metrics.mean_absolute_error(y_test, y_pred_mlr)
meanSqErr = metrics.mean_squared_error(y_test, y_pred_mlr)
rootMeanSqErr = np.sqrt(metrics.mean_squared_error(y_test, y_pred_mlr))
print('R squared: {:.2f}'.format(mlr.score(x,y)*100))
print('Mean Absolute Error:', meanAbErr)
print('Mean Square Error:', meanSqErr)
print('Root Mean Square Error:', rootMeanSqErr)
```

R squared: 80.17

Mean Absolute Error: 0.046928782854889335 Mean Square Error: 0.003716513784197383 Root Mean Square Error: 0.06096321664903668

```
y_pred_mlr= mlr.predict([[337,118,4,4.5,4.5,9.65,1],[302,102,1,2,1.5,8,0],[333,117,4
#print(x_test)
#Predicted values
print("Prediction for test set: {}".format(y_pred_mlr))
```

Prediction for test set: [0.95174021 0.54237348 0.92612384 0.59008477]

```
# create client to access our WML service
from ibm_watson_machine_learning import APIClient
client = APIClient(wml credentials)
print(client.version)
1.0.257
space_id = 'b98a964e-0e59-49d2-bef1-f71a8d8cf1af'
client.spaces.list(limit=10)
NAME
                                                        CREATED
b98a964e-0e59-49d2-bef1-f71a8d8cf1af UAEP-Deployment-Space 2022-11-18T15:47:48.722
Z
client.set.default_space(space_id)
'SUCCESS'
client.software_specifications.list()
#Upload model
software_spec_uid = client.software_specifications.get_id_by_name("runtime-22.1-py3.")
metadata = {
            client.repository.ModelMetaNames.NAME: 'Linear Regression model to predi
            client.repository.ModelMetaNames.TYPE: 'scikit-learn_1.0',
            client.repository.ModelMetaNames.SOFTWARE_SPEC_UID: software_spec_uid
published_model = client.repository.store_model(model=mlr, meta_props=metadata)
# Get model details
import json
published_model_uid = client.repository.get_model_uid(published_model)
model_details = client.repository.get_details(published_model_uid)
print(json.dumps(model_details, indent=2))
This method is deprecated, please use get_model_id()
 "entity": {
    "hybrid_pipeline_software_specs": [],
    "software_spec":
     "id": "12b83a17-24d8-5082-900f-0ab31fbfd3cb",
     "name": "runtime-22.1-py3.9"
   },
"type": "scikit-learn_1.0"
  'metadata": {
   "created_at": "2022-11-18T16:24:36.447Z",
   "id": "ec0b2afa-09de-47ad-a709-56b524cbf096",
   "modified_at": "2022-11-18T16:24:40.465Z",
   "name": "Linear Regression model to predict the admit chance",
```

```
# Create online deployment
 metadata = {
    client.deployments.ConfigurationMetaNames.NAME: "Deployment of Admit Eligibility
    client.deployments.ConfigurationMetaNames.ONLINE: {}
 }
 created_deployment = client.deployments.create(published_model_uid, meta_props=metad
###
Synchronous deployment creation for uid: 'ec0b2afa-09de-47ad-a709-56b524cbf096' star
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='10a9a42d-0d6b-42b2-8e89-d
88ce774e56b
# list all deployments
client.deployments.list()
GUID
                             NAME
STATE CREATED
10a9a42d-0d6b-42b2-8e89-d88ce774e56b Deployment of Admit Eligibility prediction mod
el ready 2022-11-18T16:48:59.382Z
5ea14582-7998-4531-805a-9cffc67841e0 Deployment of Admitb Eligibility prediction mo
del ready 2022-11-18T16:47:15.275Z
# delete old deployments
client.deployments.delete('5ea14582-7998-4531-805a-9cffc67841e0')
'SUCCESS'
client.deployments.list()
GUID
STATE CREATED
10a9a42d-0d6b-42b2-8e89-d88ce774e56b Deployment of Admit Eligibility prediction mod
el ready 2022-11-18T16:48:59.382Z
.. ....
```

7.2 Feature 2

Developed a app.py code and integrated it with the deployed model on IBM Cloud.

Respective modules are imported, flask instance is created, local database has been connected. Flask Login has been used for implementing user login.

Routes for home page and sign up page has been implemented.

```
geapp.route('/')
def home():
    return render_template('home.html')

def signup():
    return render_template('signup.html')

def signup():
    return render_template('signup.html')

def register():
    from model import user
    email=request.form.get('email')
    username=request.form.get('username')
    password=request.form.get('password')

User=user.query.filter_by(username=username).first()
    if User:
        flash('Username Already exists')
        return redirect(url_for('signup'))

new_user=user(username=username, email=email, password=generate_password_hash(password,method='sha256'))
db.session.add(new_user)
db.session.commit()
return redirect(url_for('login'))
```

Routes for login page and dashboard page has been implemented

```
@app.route('/login')
def login():
    return render_template('login.html')
@app.route('/login', methods=['POST'])
def logging_in():
   username=request.form.get('username')
    password=request.form.get('password'
    remember= True if request.form.get('Remember user') else False
    User=user.query.filter_by(username=username).first()
    if User and check_password_hash(User.password, password):
        return redirect(url_for('dashboard',current_id=User.id))
        flash('Please Check your credentials and try again')
return redirect(url_for('login'))
@app.route('/dashboard/<int:current_id>')
@login_required
def dashboard(current id):
    students=student.query.filter_by(user_id=current_id)
    return render_template('dashboard.html', collections=students, name=current_user.username)
```

Routes for add profile and edit profile has been implemented with the help of WTforms.

```
@app.route('/add_profile',methods=['GET','POST'])
def add_profile():
   form=profileform()
   if form.validate_on_submit():
      students=student(name=form.name.data, place=form.place.data, number=form.number.data, occupation=form.occupat
       value=current user.id
      students.user id=value
       db.session.add(students)
      db.session.commit()
       return redirect(url_for('dashboard',current_id=value))
   return render_template('add_profile.html', form=form)
@app.route('/edit_profile/<int:student_id>',methods=['GET','POST'])
def edit_profile(student_id):
   form=Editprofileform()
   studente=student.query.filter_by(id=student_id).first()
   if form.validate_on_submit():
       studente.name=form.name.data
      studente.place=form.place.data
       studente.number=form.number.data
       studente.occupation=form.occupation.data
       db.session.add(studente)
       db.session.commit()
       return redirect(url_for('dashboard',current_id=current_user.id))
    return render_template('edit_profile.html', form=form)
```

Routes for delete profile and predict has been implemented.

The deployed model has been used for prediction purposes with the help of the API key.

```
@app.route('/delete_profile/<int:student_id>')
def delete_deck(student_id):
     studente=student.query.filter_by(id=student_id).first()
     db.session.delete(studente)
     db.session.commit()
    return redirect(url_for('dashboard', current_id=current_user.id))
@app.route('/predict',methods=['GET','POST'])
def predict():
     if request.method=='POST':
         arr = []
for i in request.form:
          val = request.form[i]
arr.append(float(val))
          API_KEY = "gw4WhNuTcyBo4ywiRVKzN3nLK8TarnbgkpRyf_kQUgFe'
          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"]
          # NOTE: manually define and pass the array(s) of values to be scored in the next line
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/10a9a42d-0d6b-42b2-8e8theaders={'Authorization': 'Bearer' + mltoken}).json()
```

Route for logout page has been implemented.

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

value=round(result[0][0]*100,2)

if value>70:

return render_template('predict.html',

pred='High chance of admission.\nChance of admit is {} percent'.format(value))

else:

return render_template('predict.html',

pred='Low chance of admission.\nChance of admit is {} percent'.format(value))

else:

return render_template('predict.html')

else:

return render_template('predict.html')

@app.route('/logout')

@login_required

def logout():

logout_user()

return redirect(url_for('home'))

if __name__=='__main__':

app.run(debug=True)
```

7.3 Database Schema

Two Tables are used. User table and student table.

User table is used for implementing user-registration and user-login.

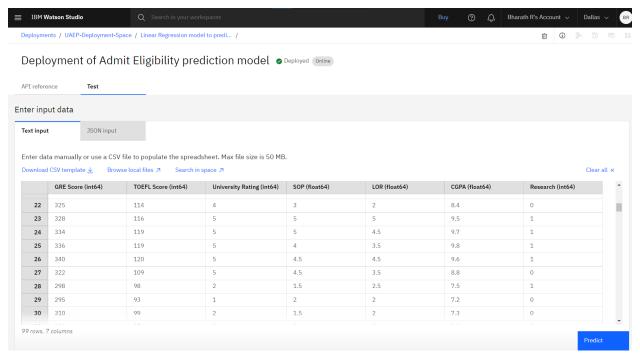
Student table is used for storing profile information.

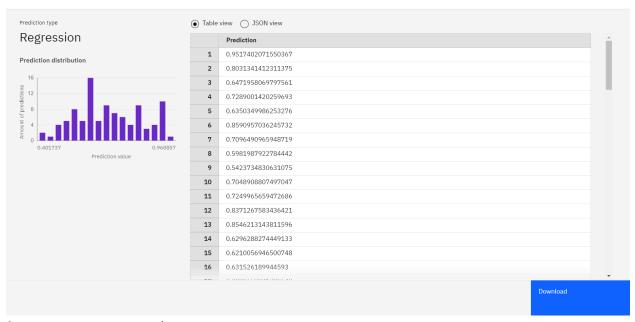
Tables (2)

Name	Туре	Schema					
student		CREATE TABLE "student" ("id" INTEGER, "name" TEXT NOT NULL, "place" TEXT NOT NULL, "number" TEXT NOT NULL, "occupation" TEXT NOT NULL, "user_id" INTEGER, PRIMARY KEY("id"), FOREIGN KEY("user_id") REFERENCES "user"("id"))					
id	INTEGER	"id" INTEGER					
name	TEXT	"name" TEXT NOT NULL					
place	TEXT	"place" TEXT NOT NULL					
number	TEXT	"number" TEXT NOT NULL					
occupation	TEXT	"occupation" TEXT NOT NULL					
user_id	INTEGER	"user_id" INTEGER					
user		CREATE TABLE "user" ("id" INTEGER, "username" TEXT NOT NULL UNIQUE, "email" TEXT NOT NULL UNIQUE, "password" TEXT NOT NULL, PRIMARY KEY("id"))					
id	INTEGER	"id" INTEGER					
username	TEXT	"username" TEXT NOT NULL UNIQUE					
email	TEXT	"email" TEXT NOT NULL UNIQUE					
password	TEXT	"password" TEXT NOT NULL					

8. Testing

8.1. Test Cases





8.2. User Acceptance Testing Defect Analysis:

Resolution	Severity 1	Severity 2	Severity 3	Severity 4	Subtotal
By Design	10	4	2	3	20
Duplicate	1	0	3	0	4
External	2	3	0	1	6
Fixed	11	2	4	20	37
Not Reproduced	0	0	1	0	1
Skipped	0	0	1	1	2
Won't Fix	0	5	2	1	8
Totals	24	14	13	26	77

Testcase Analysis:

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

9. Results

Performance Metrics

Performance metrics appropriate for linear regression are evaluated.

R squared: 80.17

Mean Absolute Error: 0.046928782854889405 Mean Square Error: 0.0037165137841973857 Root Mean Square Error: 0.0609632166490367

10. Advantages and Disadvantages

Advantages:

- The solution implemented is easy to use and follows a reliable algorithm.
- It helps aspirants an idea about where they stand now and their chances of admission to their list of dream colleges.
- It is fast, efficient, and avoids data redundancy.

Disadvantages:

- Inaccurate results can occur if the input information given is incorrect.
- The initial process can be time-consuming for some users.

11. Conclusion

Therefore, a university admit eligibility predictor is built and implemented with web application and deployed in IBM cloud. The prediction is done using linear regression in python. The web application is built using Flask. User login functionality is also implemented in web site.

12. Future Scope

- The list of universities can be increased with the inclusion of various fields of study.
- The number of countries the predictor system covers can also be increased.

13. Appendix Source Code app.py:

```
from flask import Flask,request,redirect,url_for,flash,render_template
    from flask_login.utils import login_required
    from flask_sqlalchemy import SQLAlchemy
    from werkzeug.security import check_password_hash, generate_password_hash
     from flask_login import LoginManager, login_user, current_user, logout_user
     import numpy as np
    import json
10 import requests
12 app=Flask(__name__)
14 app.secret_key='Secret key'
app.config['SQLALCHEMY_DATABASE_URI']='sqlite:///database2.sqlite3'
16 db=SQLAlchemy()
17 db.init_app(app)
    app.app_context().push()
    from model import *
22 login_manager=LoginManager()
23 login_manager.login_view='login'
24 login_manager.init_app(app)
26 @login_manager.user_loader
27 def load_user(user_id):
        return user.query.get(int(user_id))
```

```
@app.route('/')
def home():
    return render_template('home.html')

@app.route('/signup')
def signup():
    return render_template(signup.html')

@app.route('/signup',methods=['POST'])
def register():
    from model import user
    email-request.form.get('email')
    username-request.form.get('username')
    password-request.form.get('username).first()

if User:
    flash('Username Already exists')
    return redirect(url_for('signup'))

new_user-user(username-username, email-email, password-generate_password_hash(password,method-'sha256'))
db.session.add(new_user)
db.session.commit()
    return redirect(url_for('login'))

@app.route('/login')
def login():
    return render_template('login.html')

@app.route('/login', methods=['POST'])
def loging_in():
    username-request.form.get('username')
```

```
@app.route('/login', methods=['POST'])
 def logging_in():
     remember= True if request.form.get('Remember user') else False
         flash('Please Check your credentials and try again')
@login required
    students=student.query.filter_by(user_id=current_id)
     return render_template('dashboard.html', collections=students, name=current_user.username)
@app.route('/add_profile',methods=['GET','POST'])
def add profile():
    form=profileform()
     if form.validate on submit():
        students=student(name=form.name.data, place=form.place.data, number=form.number.data, occupation=form.occupation.data)
         value=current user.id
         students.user id-value
         db.session.add(students)
         db.session.commit()
    return redirect(url_for('dashboard',current_id=value))
return render_template('add_profile.html', form=form)
```

```
@app.route('/edit_profile/<int:student_id>',methods=['GET','POST'])
     def edit_profile(student_id):
         form=Editprofileform()
         studente=student.query.filter_by(id=student_id).first()
            studente.name=form.name.data
            studente.place=form.place.data
            studente.number=form.number.data
             studente.occupation=form.occupation.data
            db.session.add(studente)
            db.session.commit()
             return redirect(url_for('dashboard',current_id=current_user.id))
         return render_template('edit_profile.html', form=form)
     @app.route('/delete_profile/<int:student_id>')
110 def delete_deck(student_id):
        studente=student.query.filter_by(id=student_id).first()
         db.session.delete(studente)
         db.session.commit()
         return redirect(url_for('dashboard', current_id=current_user.id))
     @app.route('/predict',methods=['GET','POST'])
            arr = []
             for i in request.form:
                val = request.form[i]
                arr.append(float(val))
             API_KEY = "gw4WhNuTcyBo4ywiRVKzN3nLK8TarnbgkpRyf_kQUgFe"
             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"]
```

Forms.py:

```
from flask_wtf import FlaskForm
from wtforms import StringField, SubmitField, BooleanField
from wtforms.validators import DataRequired

class profileform(FlaskForm):
    name = StringField('Student Name', validators=[DataRequired()])
    place = StringField('Place', validators=[DataRequired()])
    number = StringField('Phone Number', validators=[DataRequired()])
    occupation = StringField('Occupation', validators=[DataRequired()])

submit = SubmitField('Add')

class Editprofileform(FlaskForm):
    name = StringField('Student Name', validators=[DataRequired()])
    place = StringField('Place', validators=[DataRequired()])
    number = StringField('Phone Number', validators=[DataRequired()])
    occupation = StringField('Occupation', validators=[DataRequired()])
    submit = SubmitField('Edit')
```

Templates:

Add_profile.html:

```
{% extends "layout.html" %}
{% block body %}
<div class="col-xs-12 col-sm-4 col-lg-3">
    <form method="post">
        {{ form.csrf_token }}
        <div class="form-group">
            {{ form.name.label }}
            {{ form.name(size=60, class="form-control", placeholder="Name") }}
        </div>
        <div class="form-group">
            {{ form.place.label }}
            {{ form.place(size=60, class="form-control", placeholder="Name") }}
        </div>
        <div class="form-group">
            {{ form.number.label }}
            {{ form.number(size=60, class="form-control", placeholder="Name") }}
        </div>
        <div class="form-group">
            {{ form.occupation.label }}
            {{ form.occupation(size=60, class="form-control", placeholder="Name") }}
        </div>
        <button type="submit" class="btn btn-default">Submit/button>
    </form>
</div>
```

Afterlogin.html:

```
h3 {text-align: center;}
         h5 {text-align: center;}
         div {text-align: center;}
         .flex-container{display: flex;justify-content: space-around}
<h3>Hello this is UAEP, where your LIVES are made SIMPLER!!!!</h3>
<h5>If u are a new user, please add your profile using the Add Profile option...</h5>
<h5>If u want to delete your profile please use delete profile option...</h3>
<h5>If u are an exixting user and want to edit your profile, use the Edit Profile option...</h5>
<h5>If u want to start with prediction process, use the Predict option...</h5>
<div>
          <a class="btn btn-primary" href="{{url_for('add_profile')}}" role="button">Add Profile</a>
         \label{lem:continuous} $$\arclass="btn-primary" href="{\{url\_for('predict')\}}" role="button">Predict</a>
     <div class="flex-container">
         <div class="row">
              <div class="col-lg-6 mb-4">
                  {% for collection in collections %}
                   <div class="card bg-warning" style="width: 18rem;" >
                       <div class="card-body">
                          <h5 class="card-title">Profile Information</h5><br>
                          <h6 class="card-subtitle mb-2 text-muted">Student Name: {{collection.name}}</h6>
                           <h6 class="card-subtitle mb-2 text-muted">Place: {{collection.place}}</h6>
                           <h6 class="card-subtitle mb-2 text-muted">Phone Number: {{collection.number}}</h6>
                           \label{lem:constraint} $$\an href="{\{url\_for('edit\_profile',student\_id=collection.id)\}}"$ class="card-link"> Edit Profile</a> $$\arbright = (a) $$\arbrigh
                          <a class="btn btn-danger" href="{{url_for('delete_deck',student_id=collection.id)}}" role="button">Delete</a>
                   {% endfor %}
```

Home.html:

```
{% extends "layout.html" %}
{% block body %}
<div class="col-xs-12 col-sm-4 col-lg-3">
   <form method="post">
       {{ form.csrf_token }}
        <div class="form-group">
            {{ form.name.label }}
            {{ form.name(size=60, class="form-control", placeholder="Name") }}
        <div class="form-group">
            {{ form.place.label }}
            {{ form.place(size=60, class="form-control", placeholder="Name") }}
        </div>
        <div class="form-group">
            {{ form.number.label }}
            {{ form.number(size=60, class="form-control", placeholder="Name") }}
        <div class="form-group">
            {{ form.occupation.label }}
            {{ form.occupation(size=60, class="form-control", placeholder="Name") }}
        </div>
        <button type="submit" class="btn btn-default">Submit</button>
    </form>
</div>
{% endblock %}
```

Predict.html:

```
{% extends "layout.html" %}
{% block body %}
   div {text-align: center;}
</style>
<h4>Enter your details to predict chance of admission:::</h4>
<div>
    <form action="{{ url_for('predict')}}" method="post">
        <div>
           <label for="gre_score">GRE:</label>
           <input type="text" id="gre_score" name="gre" placeholder="gre_score">
        <br>
            <label for="toefl_score">TOEFL:</label>
            <input type="text" id="toefl_score" name="toefl" placeholder="toefl_score">
       <div>
            <label for="university_rating">RATING:</label>
            <input type="text" id="university_rating" name="rating" placeholder="university_tating">
        <br>
           <label for="sop_score">SOP:</label>
            <input type="text" id="sop_score" name="sop" placeholder="sop_score">
            <label for="lor_score">LOR:</label>
            <input type="text" id="lor_score" name="lor" placeholder="lor_score">
        <div>
           <label for="cgpa_score">CGPA:</label>
            <input type="text" id="cgpa_score" name="cgpa" placeholder="cgpa_score">
        <br>
             <label for="research_score">RESEARCH:</label>
             <input type="text" id="research_score" name="research" placeholder="research_score">
        </div>
        <br>
         <div>
            <button type="submit">Enter</button>
    </form>
    <br>
     <h4>{{ pred }}</h4>
{% endblock %}
```

Dashboard.html:

```
1  {% extends "layout.html" %}
2  {% block body %}
3  <h2>Welcome {{name}}</h2>
4  {% include "afterlogin.html" %}
5  {% endblock %}
```

Edit profile.html:

```
{% extends "layout.html" %}
 {% block body %}
 <div class="col-xs-12 col-sm-4 col-lg-3">
    <form method="post">
        {{ form.csrf_token }}
        <div class="form-group">
            {{ form.name.label }}
            {{ form.name(size=60, class="form-control", placeholder="Name") }}
        </div>
        <div class="form-group">
            {{ form.place.label }}
            {{ form.place(size=60, class="form-control", placeholder="Name") }}
        <div class="form-group">
            {{ form.number.label }}
            {{ form.number(size=60, class="form-control", placeholder="Name") }}
        </div>
        <div class="form-group">
            {{ form.occupation.label }}
            {{ form.occupation(size=60, class="form-control", placeholder="Name") }}
        <button type="submit" class="btn btn-default">Submit</button>
    </form>
</div>
 {% endblock %}
```

Layout.html:

Login.html:

```
{% extends "layout.html" %}
{% block body %}
<h2>Login to see dashboard</h2>
   <div class="container">
       {% with messages = get_flashed_messages() %}
       {% if messages %}
           <div class="notification is-danger">
             {{ messages[0] }}
      {% endif %}
       {% endwith %}
      <form action="/login" method="POST">
         <div class="form-group">
              <input type="text" name="username" class="form-control" placeholder="username" required><br>
         <div class="form-group">
              <input type="password" name="password" class="form-control" placeholder="password" required><br>
           <div class="form-group">
              <label class="checkbox">
                   <input type="checkbox">
                   Remember user
              </label>
           </div>
           <div class="form-group">
              <br><button class="btn btn-primary" type="submit">submit</button>
       </form>
```

Signup.html:

```
{% extends "layout.html" %}
{% block body %}
<h1>Sign Up</h1>
   <div class="container">
      {% with messages = get_flashed_messages() %}
       {% if messages %}
           <div class="notification is-danger">
              {{messages[0]}} Go Back to <a href="{{ url_for('login') }}">Login Page</a>
       {% endif %}
       {% endwith %}
       <form action="/signup" method="POST">
           <div class="field">
              <input type="text" name="username" class="form-control" placeholder="username" required><br>
           <div class="field">
             <input type="email" name="email" class="form-control" placeholder="email" required><br>
           <div class="field">
              <input type="password" name="password" class="form-control" placeholder="password" required><br>
           <div class="form-group">
             <button class="btn btn-primary" type="submit">submit</button>
    </div>
{% endblock %}
```

Model.py (schema for local sql database using sqlalchemy):

```
from sqlalchemy.orm import relationship

from app import db

from flask_login import UserMixin

class user(db.Model,UserMixin):
    id=db.Column(db.Integer, primary_key=True)
    username=db.Column(db.String, unique=True, nullable=False)
    email=db.Column(db.String, unique=True, nullable=False)
    password=db.Column(db.String, nullable=False)

class student(db.Model):
    id=db.Column(db.Integer, primary_key=True)
    name=db.Column(db.String, nullable=False)

place=db.Column(db.String, nullable=False)
    number=db.Column(db.String, nullable=False)
    occupation=db.Column(db.String, nullable=False)
    user_id=db.Column(db.Integer, db.ForeignKey(user.id), nullable=False,)

user_id=db.Column(db.Integer, db.ForeignKey(user.id), nullable=False,)
```

GitHub & Project Demo Link

Github repo link:

https://github.com/IBM-EPBL/IBM-Project-19834-1659707682

Project Demo Link:

https://www.youtube.com/embed/6sa_H0Jmewc