

SOMAIYA VIDYAVIHAR

K J Somaiya Institute of Engineering and Information Technology

An Autonomous Institute affiliated to University of Mumbai NBA Accredited 3 Programs (Computer Engineering, Electronics & Telecommunication Engineering & Electronics Engineering) Accredited by NAAC and NBA, Approved by AICTE, New Delhi

Synopsis of Mini Project On

University Recommendation System

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Choice based Credit Grading System - SEM VIII (LY - IT)

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1. DEPARTMENT OF INFORMATION TECHNOLOGY **CERTIFICATE**

This is to certify that following students:
Roll number
Akash Anand Prabhudesai 51
have submitted Project Report on " <i>University Recommendation System</i> " as the fulfillment for the requirement of Final Year of Engineering (8 th Semester) in B. Tech L.Y Information Technology under my guidance during the academic year 2021-2022.
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Date of Examination:
Signature of Internal Examiner Signature of External Examiner
2 Table of Contents

1. ABSTRACT...... 4

2. INTRODUCTION	4
3. BACKGRUND & OBJECTIVE	4
4. DATASET	4
1. Graduate dataset	5
2. Undergraduate dataset	5
3. Data Preprocessing	6
4. Exploratory data Analysis	7
5. ANALYSIS & METHODOLOGY	8
6. IMPLEMENTATION	9
7. RESULTS	10
8. DATA PREPROCESSING	10
9. RECOMMENDATION SYSTEM	12
10. CONCLUSION	14
11. REFERENCES	15

1. ABSTRACT:

admission seekers, which can help students to choose best graduate university matching their academic profile. Here I have used a different data mining technique to transform database of students of relevant information into a universal database format using academic data of successful students who have already got opportunity to study abroad. After that I have developed a machine learning algorithm which can calculate similarity between training and test data based on weighted scores. I have used K-nearest Neighbor algorithm and feature-weighted algorithm for calculating top N similar users for the test users and recommend Top K universities to users from N similar users.

2. INTRODUCTION:

Many students who wants to pursue higher studies apply different universities with their academic profile as well as standardized test scores such as SAT, GRE, TOEFL, and IELTS. Institutions take in the students who are suitable candidates based on their academic profile, standardized test scores. But in this entire process university selection is the most crucial & tedious step for applying to graduate studies. Some of them succeed and get admission into their desired programs in desired universities, but some are not because of the academic level of colleges which they have applied. To resolve this problem of not getting admission because of applications, even though students have good academic profile, I have developed this recommendation system. In this project, the knowledge acquired from the database of successful applicants is used to predict the schools with various data mining techniques. This data will be modeled into machine learning algorithms to predict the universities and their acceptance rate for the given user academic details.

3. BACKGROUND & OBJECTIVE:

For an aspiring student who wants to apply for higher studies in other countries, university selection process is a challenging task as lot of different criteria need to consider during application process based on individual's requirement. Some of them succeed and get admission into their desired programs in desired universities, but some are not because of the academic level of colleges, which they have applied.

This problem can be addressed by modeling a recommender system based on various classification algorithms. In this project based on the student data set and the student profile who is looking for the admit, various models will be trained and a list of 10 best universities will be suggested such that it maximizes the chances of a student getting admit from that university list.

4. DATASET:

The first step in building any recommendation system is the identification of the data set. In order to build the classification model for the recommender system, this data has to be organized with appropriate labels. This core data for the application process is not readily available on the internet for direct consumption. However, this whole approach is based on making maximum use of the

available information. The graduate student data was scraped from the following websites

4.1. Graduate Student Dataset:

For Graduate Student data, we scraped <u>www.thegradcafe.com</u> website. About 271807 rows of raw student data was obtained as a result of web scraping. Each sample corresponds to the profile of a student. We have got 1949 html pages of the data and need to change it into CSV files.

```
import matplotlib.pyplot as plt
import requests
import urllib.request
from IPython.core.debugger import Tracer
url_form = "http://thegradcafe.com/survey/index.php?q-u%2A&t-a&pp-250&o-d&p-{0}"
DATA_DIR = './WebScraped_data/html/'
           --
                main
    name
    for i in range(1691, 1948):
        url = url_form.format(i)
        handle = urllib.request.urlopen(url)
        html = handle.read()
        html = html.decode('utf8')
        #r = requests.get(url)
        fname = "{data_dir}/{page}.html".format(data_dir=DATA_DIR,page=str(i))
        with open(fname, 'wb') as f:
            f.write(html.encode('UTF-8'))
        print("getting {0}...".format(i))
getting 1775...
getting 1776...
getting 1777...
getting 1778...
getting 1779...
getting 1780...
getting 1781...
```

After scraping the final data will look like



The list of attributes are made as dataset for pre-process cleansing. For graduate students the dataset consists of University Name, Major, Degree, Season, Decision, Decision Method, Decision Date, Undergraduate GPA, Is New GRE Verbal, GRE Quant, GRE Writing, Status, Postdate Comments, Research Experience, Recommendations and Undergraduate GPA. For Under graduate student's dataset consists of Student profile and SAT scores.

5

4.2. Under Graduate student dataset:

Under Graduate student data is taken from the College rank score card website https://collegescorecard.ed.gov/data/. The data before cleaning looked like below.

INJUNITIO	OPEIO	OPEIDS	INSTAM	CITY	STABBIN	2P	ACCREDAGE	DASTURE.	MPCUM.	SCH_DEG	HCMQ	MMUN	NUMBRANCI PREDDEG	HIGHOEG	CONTROL	ST_FPS	REGICAL
200694	100300	1000	Alabama A.	& Normal	AL.	35762	MULL	PULL	MUU.	NULL	NUL		1 1	1	4	1	1
200668	105300	1052	University	d birmingham	AL.	35294-0110	MULL.	PULL	MUU.	NULL	NUL		1 1	1	6	1	1
200690	2903400	21034	Amridge Un	vi Morragomen	AL.	36117-0910	MULL	PAUL	MUU.	NULL	MUU.		1 1	3	6	2	1
100706	105500	1055	University	d Hustoville	AL.	35899	MULL	PRULL	MUSE	NULL	MULL		1 1	3	4	1	1
100734	100500	1005	Alabama St.	a Montgomer	AL.	36304-0273	MULL	PRULL	MUSE	NULL	MULL		1 1	3	4	1	1
500754	105100	1051	The Univers	of Tuncalouss	AL.	35487-0366	MULL	PRULL	MUSE	MULL	MUUL		1 1)	4	1	1
500760	100790	1007	Central Alah	bi Alexander ()	AL.	35000	MULL	PRUILL	MULL	NULL	MUU.		1 1	2	1	1	1
100812	100600	1006	Athens Stat	se Atheno	AL.	39611	MULL	PAUL	MULL	NULL	MUU.		1 1	3	3	1	1
100830	853000	8030	Auburn Uni	in Montgomen	AL.	36117-5996	MULL.	NUL.	MUU.	NULL.	NUL.		1 1	3	4	1	1
100858	100900	1009	Auburn Uni	in-Aubum	AL.	36849	MULL	NULL.	MUU.	NUL.	NUL.		1 1	3	4	1	1
100937	101100	1012	Birmingham	Sirmingham	AL.	35254	MULL	NULL.	MUU.	NULL	NUU.		1 1	3	8	2	1

4.3. <u>Data Preprocessing:</u>

In order to use the obtained data for our analysis, we need to do the preprocessing and cleansing, as there are lots of anomalies in the dataset. For this we use pandas and numpy frameworks.

Cleansing the data was done by

- Removing the irrelevant columns by using the drop column feature
- Filling the null values with the appropriate value or deleting the row containing null values. Removing the spaces in the data and reducing the size of the dataset.

In our graduate dataset, The GRE scores were also cleansed since they contained scores of both old and new versions of the examination. Similarly, the GPA scores available were based on different point systems, so all the GPA scores were uniformly scaled to 4 point scale by using normalize functions.

$$X_{\text{normalized}} = (X - X_{\text{minimum}}) / (X_{\text{maximum}} - X_{\text{minimum}})$$

Where x is the value of the GPA

4.4. Exploratory data Analysis:

Exploratory data Analysis is a technique, which employs a variety of techniques. It consists of various techniques like below.

- 1.Plotting raw data
- 2. Plotting simple statistics such as mean, standard deviation, etc.
- 3. Positioning such plots, so as to maximize our natural pattern recognition.

This is the description of the data.

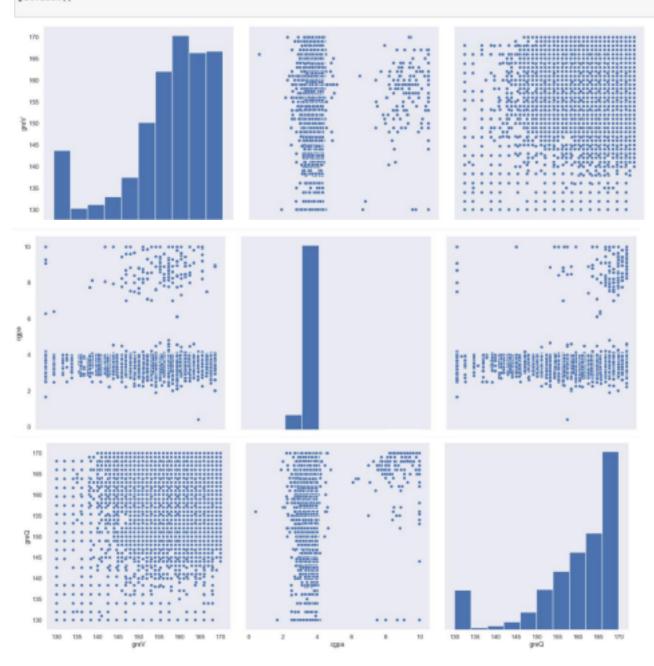
data.describe()

	decdate_ts	cgpa	greV	greQ	greA	gre_subject	post_timestamp
count	6.145400e+04	55589.000000	61474.000000	61474.000000	61474.000000	7175.000000	6.147400e+04
mean	1.431551e+09	3.715970	231.556333	248.826447	4.144757	796.411150	1.431763e+09
std	9.540728e+07	0.506153	174.575147	208.551820	1.111126	122.305977	8.079993e+07
min	-1.000000e+00	0.400000	130.000000	130.000000	0.000000	310.000000	1.263283e+09
25%	1.363244e+09	3.520000	155.000000	157.000000	3.500000	710.000000	1.363417e+09
50%	1.426662e+09	3.750000	161.000000	164.000000	4.000000	800.000000	1.427094e+09
75%	1.490771e+09	3.900000	167.000000	170.000000	5.000000	890.000000	1.491030e+09
max	1.360120e+10	9.990000	800.000000	800.000000	6.000000	990.000000	1.562569e+09

The Processed data will look like below.



Exploratory data analysis: pair plots of GRE verbal, GRE Quantitative and GPA.



Undergraduate Data EDA:

In Undergraduate data, we have taken below few rows of data like Institution name, city, tuition Fees, Sat Score, Admission rate, Debt and Men Ratio.

```
INSTNM', 'CITY', 'STABBR' ,'TUITIONFEE_OUT', 'SAT_AVG_ALL', 'ADM_RATE_ALL',
'DEBT MDN SUPP', 'UGDS MEN'])
```

This data will be used for training the model and test data as SAT score and Maximum tuition fees.

5. ANALYSIS & METHODOLOGY

Here I have used Knowledge based recommendation System where User inputs are taken into account and compare with the training data.

For Graduate University Recommendation I have used Case based knowledge recommendation as it will take the User inputs and compare with trained data.

For Undergraduate Recommendation System, I have used Constraint based Knowledge recommendation system where user inputs taken into account as constraints and based on the constraints I compared with trained data.

I used two different models like K-Nearest Neighbors for Graduate data and Feature weighted algorithms for Undergraduate data.

K Nearest Neighbor:

In KNN, the trained data is compared with test data and distances are calculated using Euclidean distance. It then classifies an instance by finding its nearest neighbors and recommend the top n nearest neighbor universities.

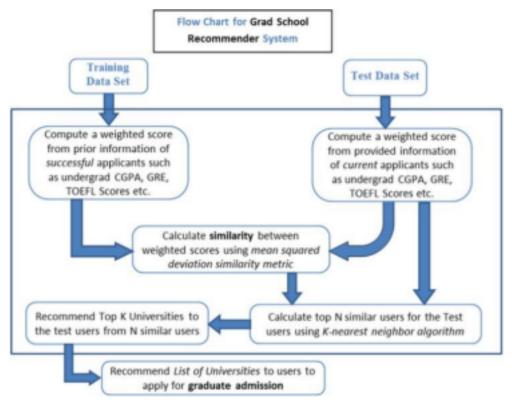
Algorithm is stated as below.

Input: undergraduate university, department, CGPA, GRE Scores of User

- 1. Initialize the value of k
- 2. For getting recommendation, iterate from 1 to number of trained data
- 3. Calculate distance between test data and each row in the trained data.
- 4. Sort the distances in ascending order
- 5. Get top k rows and recommend to the user

Output: highly recommended N Outgoing University analyzing Universal Table of Previous Successful Students

Flowchart of the graduate recommendation System:



6. IMPLEMENTATION

Training data for the KNN algorithm:

	univName	cgpa	greV	greQ	greA
14	Ohio State University	4.00	150.0	166.0	3.0
17	Texas A&M University	3.57	157.0	151.0	5.5
46	University Of California, Irvine	3.66	155.0	167.0	4.0
64	Boston University	3.10	161.0	157.0	4.0
203	Oregon State University	3.38	154.0	170.0	4.0

```
def knn(trainingSet, testInstance, k):
   print(k)
   distances = {}
   sort = {}
   length = testInstance.shape[1]
    for x in range(len(trainingSet)):
        dist = euclideanDistance(testInstance, trainingSet.iloc[x], length)
        distances[x] = dist[0]
    sorted_d = sorted(distances.items(), key=lambda x: x[1])
    neighbors = []
    for x in range(k):
       neighbors.append(sorted_d[x][0])
    classVotes - {}
    for x in range(len(neighbors)):
       response = trainingSet.iloc[neighbors[x]][-1]
       if response in classVotes:
           classVotes[response] - 1
           classVotes[response] = 1
    sortedVotes = sorted(classVotes.items(), key-lambda x: x[1], reverse-True)
    return(sortedVotes, neighbors)
```

Implementation of Feature weighted algorithm for Undergraduate universities:

The weightage of all the features are taken and find the similarity score. Based on the similarity score, the universities with highest similarities will be recommended to student. Suppose w1, w2 are weights and f1 and f2 are features the similarity is calculated by formula Similarity

$$score = w1* f1+w2*(1-f2)$$

Algorithm is stated as below.

Input: SAT Score and Maximum tuition fees of User

- 1. For getting recommendation, iterate from 1 to number of trained data
- 2. Find the rows in the training data similar to the user provided SAT score and max tuition fees.
- 3. Calculate the weightage of both the attributes and calculate the score as acceptance rate
- 4. Sort the distances in ascending order
- 5. Get top k rows and recommend to the user

Output: Top 5 Recommended Universities

6. RESULT

For graduate University recommendation of KNN algorithm, for the input test = [145, 156, 4, 3.8], The result is provided as:

```
k = 7
result, neigh = knn(processed_data, test, k)
list2 = []
for i in result:
    list1.append(i[0])
    list2.append(i[1])
for i in list1:
    print(i)
University Of Colorado, Boulder
University Of Florida
University Of Arizona
University Of Pennsylvania (UPenn)
Syracuse University
University Of Texas At Austin
Emory University
from sklearn.neighbors import KNeighborsClassifier
neigh = KNeighborsClassifier(n_neighbors=5)
neigh.fit(processed_data.iloc[:,0:4], data['univName'])
print(neigh.predict(test))
['Syracuse University']
```

For Undergraduate Universities recommendation, weighted algorithm outputs:

Input of the data for SAT Score and Maximum Tuition Fees:

```
please input your maximum tuition you can accept:10000

('Fort Valley State University': <_main_.CollageInfo object at 0x119e300f0x, 'Bemidji State University': <_main_.CollageInfo object at 0x119e30f08x, 'Bemidji State University': <_main_.CollageInfo object at 0x119e30f08x, 'Bemidji State University': <_main_.CollageInfo object at 0x119e70f08x, 'Southwest Minnesota State University': <_main_.CollageInfo object at 0x119e70f08x, 'Delta State University': <_main_.CollageInfo object at 0x119e70f08x, 'Delta State University': <_main_.CollageInfo object at 0x119e70f08x, 'Mississippi Valley State University': <_main_.CollageInfo object at 0x119e70f08x, 'United States Merchant Marine Academy': <_main_.CollageInfo object at 0x119ec19b0x, 'Dickinson State University': <_main_.CollageInfo object at 0x119ed30f0x, 'Minot State University': <_main_.CollageInfo object at 0x119ed30f0x, 'Central State University': <_main_.CollageInfo object at 0x119ec080f0x, 'Youngstown State University': <_main_.CollageInfo object at 0x119f05f0f0x, 'Paul Quinn Collage': <_main_.CollageInfo object at 0x119f05f0f0x, 'Brigham Young University': <_main_.CollageInfo object at 0x119f0f0f0f0x, 'Brigham Young University': <_main_.CollageInfo object at 0x119f0f0f0f0x, 'Brigham Young University': <_main_.CollageInfo object at 0x119f0f0f0f0x, 'Brigham Young University': <_main_.CollageInfo object at 0x119f0f
```

12

Output:

college : Brigham Young University-Idaho score: 0.8108348810996646 city:Rexburg state: ID rank: 178 tuition:3920.0 sat:1036.0 AC:0.95830221558376 debt:668 Mal:0.4236 college : Mississippi Valley State University score: 0.7441481637097037 city:Itta Bena state:MS rank: 1243 tuition:6116.0 sat:825.0 AC:0.84452975047984 debt:185 Mal:0.4146 .0 college : Alcorn State University score: 0.6810460877865222 city:Alcorn State rank:1127 tuition:6546.0 sat:892.0 AC:0.78440808469682 Mal:0.362 bt:19000.0 college: The University of Texas of the Permian Basin score: 0.6696832133488956 rank: 900 tuition:6958.0 sat:953.0 AC:0.81484315225707 city:Odessa state:TX debt:838 Mal:0.4428 college : Delta State University score: 0.637365259808179 city:Cleveland state:MS rank: 997 tuition:6418.0 sat:1028.0 AC:0.89407744874715 debt:1239 Mal:0.4138

13

7. CONCLUSION:

This project helps students in the decision making of the universities in which they apply. The data of the previous successful applicants can be taken into account. The data from the academic records of applicants is very important for the admission seekers in foreign. In this research, I have developed a technique of using those academic records of successful applicants for making school recommender system, which can help the current admission seekers. At first, I calculate similarity between training and test data set based on weighted scores. The weighted scores are calculated from prior information of successful applicants such as undergrad CGPA, GRE, TOEFL Scores and all other relevant records found in the universal database. I have used K-nearest Neighbor algorithm for graduate universities and feature weighted algorithm for Undergraduate Universities in order to calculate top N similar users and then recommend top K universities to the users. Our proposed recommender system will recommend list of universities to applicants trying to pursue higher study abroad and assist them to apply for graduate admission in appropriate universities.

14

8. REFERENCES:

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- 2. https://www.semanticscholar.org/paper/Recommender-System-for-Graduate-Studies-in-USA-Suresh/22924fda3f293f80a3f62f32799c08d0b81a9b20
- 3. https://www.researchgate.net/publication/311758642 Graduate school recommender syste <a href="mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:mailto:m
- 4. http://jmcauley.ucsd.edu/cse258/projects/fa15/026.pdf
- 5. https://devpost.com/software/graduate-school-recommendation-system
- 6. http://paper.ijcsns.org/07 book/201801/20180111.pdf