

PROJECT REPORT

INTELLIGENT ADMISSIONS: THE FUTURE OF UNIVERSITY DECISION MAKING WITH MACHINE LEARNING

PROJECT DESCRIPTION

- University admission is the process by which students are selected to attend a college or university. The process typically involves several steps, including submitting an application, taking entrance exams, and participating in interviews or other evaluations.
- Students are often worried about their chances of admissions in university. The university admissions process for students can be demanding, but by being well-informed, prepared, and organized, students can increase their chances of being admitted to the university of their choice.
- **The project aim to predict the chances of a student getting admitted to a particular university based on certain factors the business values of this project is that it will helped students make more informed decision about which universities to apply to , and help university counsellors to better advise students on the universities they are most likely to be admitted to the university.**
- The ability to accurately predict the chances of university admissions can help students make more informed decision about which university to apply to, increasing the chances of begin admitted and ultimately gaining access to higher education

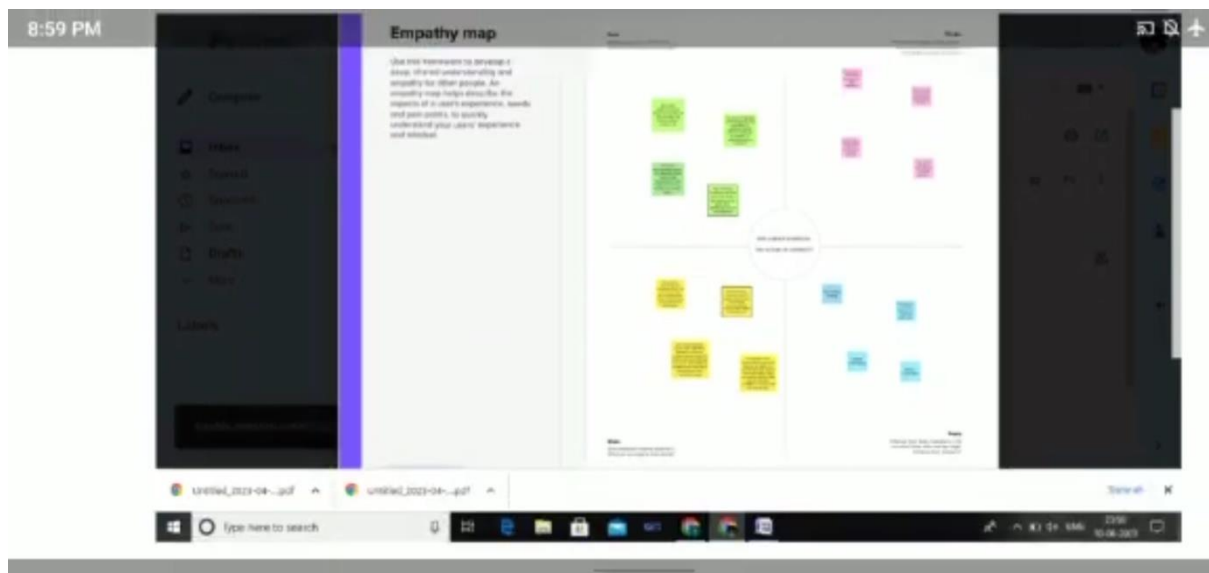
THE PURPOSE OF THE PROJECT:

- ✓ The aim of this project is to help students in short listing universities with their profiles. Machine learning algorithms are then used to train a modal on this data, which can be used to predict the chances of future applicants being admitted. With this project, students can make more informed decisions about which universities to apply to, and universities can make more efficient use of their resources by focusing on the most promising applicant.

- ✓ 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 a better idea

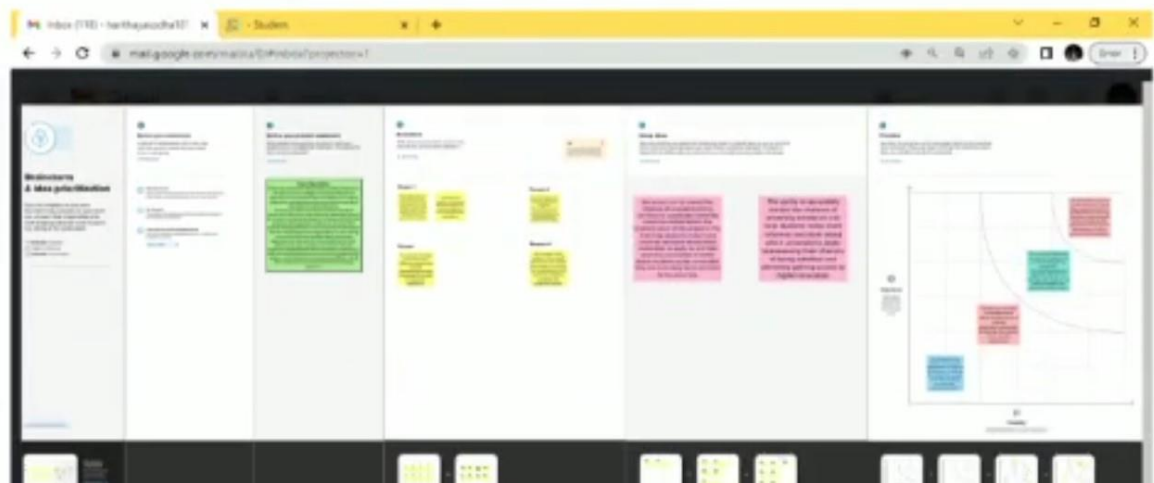
PROBLEM DEFINITION AND DESIGN THINKING:

EMPATHY MAP



BRAINSTORMING MAP

Brainstroming map:



RESULT:

```
Data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 400 entries, 0 to 399
Data columns (total 9 columns):
 #   Column              Non-Null Count  Dtype  
---  -
 0   Serial No.         400 non-null   int64  
 1   GRE Score           400 non-null   int64  
 2   TOEFL Score        400 non-null   int64  
 3   University Rating   400 non-null   int64  
 4   SOP                 400 non-null   float64 
 5   LOR                 400 non-null   float64 
 6   CGPA                400 non-null   float64 
 7   Research            400 non-null   int64  
 8   Chance of Admit     400 non-null   float64 
dtypes: float64(4), int64(5)
memory usage: 28.2 KB

0s completed at 14:20
```

Google Colab interface showing a Jupyter Notebook. The code cell contains the following Python code:

```
[3]: 6 CGPA 400 non-null float64
      7 Research 400 non-null int64
      8 Chance of Admit 400 non-null float64
      dtypes: float64(4), int64(5)
      memory usage: 28.2 KB
```

The output cell shows the result of `Data.isnull().any()`:

```
Serial No. False
GRE Score False
TOEFL Score False
University Rating False
SOP False
LOR False
CGPA False
Research False
Chance of Admit False
dtype: bool
```

The status bar indicates 84.49 GB available and the cell is completed at 14:22. The page number 4/11 is visible at the bottom.

Google Colab interface showing a Jupyter Notebook. The code cell contains the following Python code:

```
Data.describe()
```

The output cell shows the descriptive statistics for the dataset:

	Serial No.	GRE Score	TOEFL Score	University Rating	SOP	LOR	CGPA
count	400.000000	400.000000	400.000000	400.000000	400.000000	400.000000	400.000000
mean	260.500000	316.807500	107.410000	3.087500	3.400000	3.452500	8.508925
std	115.614301	11.473646	6.069514	1.143728	1.006869	0.586478	0.506317
min	1.000000	280.000000	82.000000	1.000000	1.000000	1.000000	6.800000
25%	100.750000	308.000000	103.000000	2.000000	2.500000	3.000000	8.170000
50%	260.500000	317.000000	107.000000	3.000000	3.500000	3.500000	8.610000
75%	360.250000	325.000000	112.000000	4.000000	4.000000	4.000000	9.062500
max	400.000000	340.000000	120.000000	5.000000	5.000000	5.000000	9.920000

The status bar indicates 84.49 GB available and the cell is completed at 14:22. The page number 5/11 is visible at the bottom.


```
from sklearn.metrics import accuracy_score, recall_score, roc_auc_score, confusion_matrix
print(classification_report(y_train, pred))
```

	precision	recall	f1-score	support
False	1.00	0.10	0.20	20
True	0.93	1.00	0.97	280
accuracy	0.97	0.98	0.97	300
macro avg	0.97	0.93	0.93	300
weighted avg	0.96	0.93	0.93	300

```
from sklearn.metrics import accuracy_score, recall_score, roc_auc_score
print(classification_report(y_test, pred))
```

	precision	recall	f1-score	support
False	0.00	0.00	0.00	10
True	0.88	1.00	0.94	290
accuracy			0.88	300
macro avg	0.44	0.50	0.47	300
weighted avg	0.44	0.88	0.67	300

ADVANTAGES OF INTELLIGENT ADMISSIONS:

- It helps student for making decision for choosing a right college
- Here the chance of occurrence of error is less when compared with the existing system.
- It is fast, efficient and reliable.
- Avoids data redundancy and inconsistency.
- Very user-friendly.
- Easy accessibility of data.
- Student and placement officer post are first sent to admin for approval. It keeps the system stable by not posting any spam or miscellaneous content on the web.

DISADVANTAGES OF INTELLIGENT ADMISSIONS:

- Required active internet connection.
- System will provide inaccurate results if data entered in correctly.
- May produce inaccurate results if the data is not feed properly.
- Requires an active internet connection

APPLICATIONS

- Schools and colleges
- Medical industry
- Business
- Campus

CONCLUSION

Future work will focus on implementing the proposed architecture for the education system.

FUTURE SCOPE

High level information technology skills will continue to be in demand as technology advantages and become more intergrated into every area of daily life .

APPENDIX:

```
Import numpy as np
```

```
Import pandas as pd
```

```
Import matplotlib.pyplot as plt
```

```
Impor seaborn as sns
```

```
%matplotlib inline
```

```
Data=pd.read_csv('Admission_predict.csv')
```

```
Data.info()
```

```
Data.isnull().any()
```

```
Data=data.rename(columns={'Change of Admit':'Change of Admin'})
```

```
Data.describe()
```

```
Sns.displot(data['GRE Score'])
```

```
Sns.pairplot(data=data,hue='Research',markers=['^','v'],palette='inferno')
```

```
Sns.scatterplot(x='University Rating',y='CGPA',data=data,color='Red',s=100)
```

```
Category=['GRE Score','TOEFL Score','University  
Rating','SOP','LOR','GGPA','Research','change of admit']
```

```
Color='yellowgreen','gold','lightskyblue','pink','red','purplr','orange','gray']
```

```
Start=True
```

```
For i in np.arange(4):
```

```
    Flg=plt.figure(figsize=((14,8))
```

```
    Plt.subplot2grid((4,2),(i,0))
```

```
    Data(category[2*i].his(color=color[2*i]P.bins=10)
```

```
    Plt.title(category[2*i])
```

```
    Plt.subplot2grid((4,2)'(i,1))
```

```
    Data[category[2*i+1].hist(color=color[2*i+1],bins=10))
```

```
    Plt.title(category[2*i+1])
```

```
Plt.subplots_adjust(hspace - 0.7,wspace - o.2)
```

```
Plt.show()
```

```
From sklearn.preprocessing import MinMaxScaler
```

```
Sc=MinMaxScaler()
```

```
X=sc.fit_transform(x)
```

```
X
```

```
X=data.iloc[:0:7].values
```

```
X
```

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

```
Y
```



```

From sklearn.model_selection import train_test_split

X_train,x_test,y_train,y_test = train_test_split(x,y
test_size=0.30,random_state+101)

Y_train=(y_train>0.5)

Y_train

Y_test=(y_test>0.5)

From sklearn.linear_model.logistic import LogisticRegression

Cls=Logistic Regression(random_state=0)

lr=cls.fit(x_train,y_train)

Y_pred=lr.predict(x_test)

Y_pred

Import tensorflow as tf

From tensorflow import keras

From tensorflow.keras.layers import Dense,Activation,Dropout

From tensorflow.keras.optimizers import Adam

Model=keras.Sequential()

Model.add(Dense(7,activation='relu',input_dim=7))

Model.add(Dense(7,activation='relu'))

Model.add(Dense(7,activation='relu'))

Model.add(Dense(1,activation='linear'))

Model.summary()

Model.fit(x_train,y_train,batch_size=20,epochs=100)

Model.compile(loss='binary_crossentropy',optimizer='adam',metrics=['accuracy'])

```

```
Model.fit(x_train,y_train,batch_size=20,epochs=100)

From sklearn.metrics import accuracy_score

Train_acc=model.evaluate(x_train,y_train,verbose=0)[1]

Print (train_acc)

Test_acc=model.evaluate(x_test,y_test,verbose=0)[1]

Print (test_acc)

Pred=model.predict (x_test)
```