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

1. INTRODUCTION

1.1 OVERVIEW

Intelligent admissions is a process of using machine learning algorithms to predict student admissions rate into universities.

It is a way of automating the decision-making process in university admissions. The use of machine learning algorithms in university admissions has been shown to be effective in predicting student success rates and improving the accuracy of admission decisions

Intelligent Admissions is a project that aims to use machine learning algorithms to improve the university admissions process.

The project involves using data from previous admissions cycles to train machine learning models that can predict which students are most likely to be admitted.

These models can then be used to help universities make more informed decisions about which students to admit.

The project has several potential benefits, including increased efficiency in the admissions process, improved accuracy in predicting which students are most likely to be admitted, and increased diversity in the student body

1.2 PURPOSE

The university admission prediction project can be used to help students make more informed decisions about which universities to apply to, increasing their chances of being admitted and ultimately gaining access to higher education.

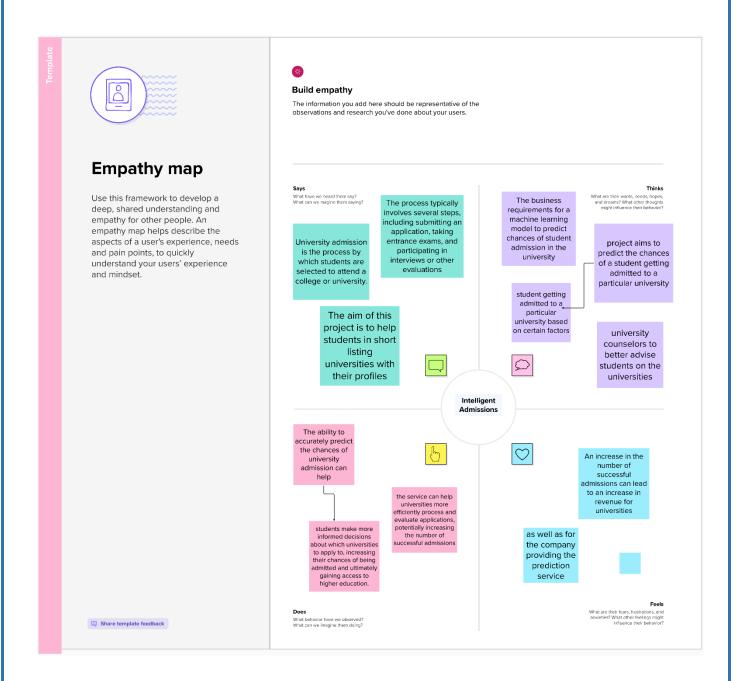
Additionally, using machine learning models to predict university admission can help universities more efficiently process and evaluate applications, potentially increasing the number of successful admissions.

An increase in the number of successful admissions can lead to an increase in revenue for universities, as well as for the company providing the prediction service.

2. PROBLEM DEFINITION AND DESIGN THINKING

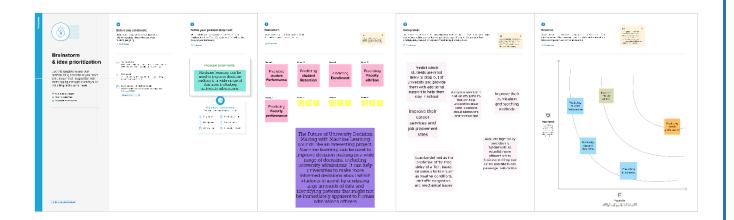
2.1 EMPATHY MAP

- ✓ Predicting student Performance
- ✓ Predicting student Retention
- ✓ Predicting Enrollment
- ✓ Predicting Faculty attrition
- ✓ Predicting Faculty performance



2.2 IDEATION AND BRAINSTORMING MAP

- ✓ Predict which students are most likely to drop out of university and provide them with additional support to help them stay in school
- ✓ Analyze student data and identify patterns that can help universities make better decisions about admissions and financial aid
- ✓ Improve their curriculum and teaching methods
- ✓ Improve their career services and job placement rates



3. RESULT

University admission process can be demanding and stressful for students.

I'm glad to hear that machine learning algorithms are being used to help students 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 applicants.

Machine learning algorithms can help students 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 applicants.

This is a great way to help students who are currently preparing or will be preparing to get a better idea about their admission chances in a particular university.

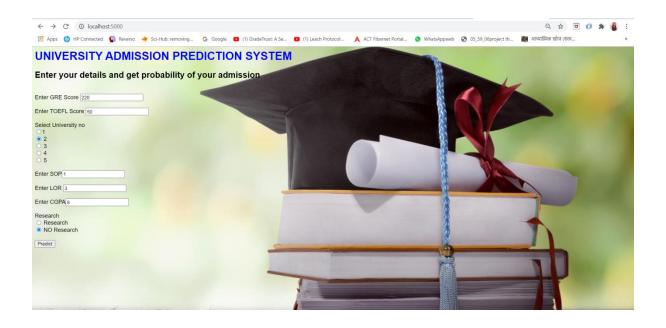
The business value of this project is that it will help students make more informed decisions about which universities to apply to, and help university counselors to better advise students on the universities they are most likely to be admitted to the university.

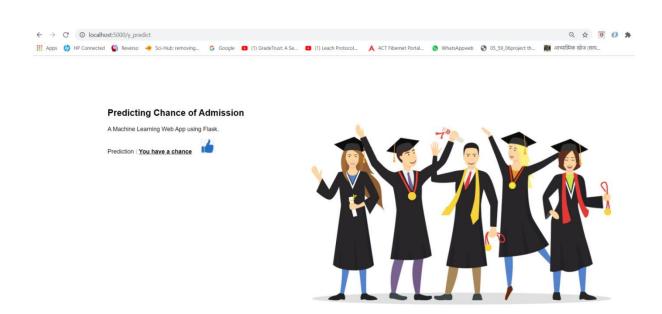
Various machine learning algorithms can be used to predict the chance of university admission with high accuracy. The ability to accurately predict the chances of university admission can

help students make more informed decisions about which universities to apply to, increasing their chances of being admitted and ultimately gaining access to higher education.

Additionally, using machine learning models to predict university admission can help universities more efficiently process and evaluate applications, potentially increasing the number of successful admissions.

An increase in the number of successful admissions can lead to an increase in revenue for universities, as well as for the company providing the prediction service.





4. ADVANTAGES AND DISADVANTAGES

4.1 ADVANTAGES

- ✓ The advantages of the university admission prediction project include helping students make more informed decisions about which universities to apply to, increasing their chances of being admitted and ultimately gaining access to higher education.
- ✓ Additionally, using machine learning models to predict university admission can help universities more efficiently process and evaluate applications, potentially increasing the number of successful admissions.
- ✓ An increase in the number of successful admissions can lead to an increase in revenue for universities, as well as for the company providing the prediction service.

4.2 DISADVANTAGES

- ✓ Some potential disadvantages of using machine learning models for university admission prediction include the possibility of bias in the data used to train the model, which could lead to unfair or discriminatory admissions decisions.
- ✓ Additionally, machine learning models may not be able to account for all relevant factors that could impact admissions decisions, such as extracurricular activities or personal essays.
- ✓ Finally, machine learning models may not be able to account for changes in admissions policies or procedures over time.

5. APPLICATIONS

The university admission prediction project can be applied in a variety of contexts, including helping students make more informed decisions about which universities to apply to, increasing their chances of being admitted and ultimately gaining access to higher education.

Additionally, using machine learning models to predict university admission can help universities more efficiently process and evaluate applications, potentially increasing the number of successful admissions.

An increase in the number of successful admissions can lead to an increase in revenue for universities, as well as for the company providing the prediction service.

6. CONCLUSION

To summarize, the university admission prediction project uses machine learning models to predict university admission and can be applied in a variety of contexts, including helping students make more informed decisions about which universities to apply to and increasing their chances of being admitted. However, there are also potential disadvantages to using machine learning models for university admission prediction, such as the possibility of bias in the data used to train the model.

7. FUTURE SCOPE

Some potential enhancements that could be made to the university admission prediction project in the future include improving the accuracy of the machine learning models used to predict university admission, as well as increasing the number of factors that are taken into account when making admissions decisions.

Additionally, it may be possible to use machine learning models to predict other aspects of the university experience, such as student retention rates or graduation rates

<u>APPENDIX</u>

SOURCE CODE

```
#!/usr/bin/env python
# coding: utf-8
# In[5]
@app.route('/')
def home():
  return render_template('Demo2.html')
# In[1]:
@app.route('/')
def home():
  return render_template('index.html')
@app.route('/y_predict',methods=['POST'])
```

```
def y_predict():
  For rendering results on HTML GUI
  #min max scaling
  min1=[290.0, 92.0, 1.0, 1.0, 6.8, 0.0]
  \max 1 = [340.0, 120.0, 5.0, 5.0, 9.92, 1.0]
  k= [float(x) for x in request.form.values()]
  p=[]
  for i in range(7):
     1=(k[i]-min1[i])/(max1[i]-min1[i])
     p.append(1)
  prediction = model.predict([p])
  print(prediction)
  output=prediction[0]
  if(output==False):
     return render_template('noChance.html', prediction_text='You Dont have a chance of
gettin')#not complete
  else:
    return render_template('chance.html', prediction_text='You have a chance of getting
admis')
  if _name_=="__main__":
     app.run(debug=False)
# In[]:
code.ipynb
#!/usr/bin/env python
# coding: utf-8
# In[6]:
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
get_ipython().run_line_magic('matplotlib', 'inline')
# In[7]:
data = pd.read_csv(r"C:\Users\vigne\Intelligent Admissions The Future of University
Decision Making with Machine Learning\Dataset\Admission_Predict.csv")
# In[17]:
data.info()
# In[18]:
data.isnull().any()
# In[23]:
data = data.rename(columns = {'Chance of Admit':'Chance of Admit'})
# In[24]:
data.describe(
# In[27]
sns.distplot(data['GRE Score'])
# In[28]:
sns.pairplot(data=data,hue='Research',markers=["^","v"],palette='inferno')
# In[29]:
sns.scatterplot(x='University Rating',y='CGPA',data=data,color='Red',s=100)
# In[15]
import pandas as pd
import numpy as np
```

```
import matplotlib.pyplot as plt
# Define the data
data = pd.DataFrame({
  'GRE Score': np.random.randint(300, 340, size=100),
  'TOEFL Score': np.random.randint(90, 120, size=100),
  'University Rating': np.random.randint(1, 6, size=100),
  'SOP': np.random.uniform(1.0, 5.0, size=100),
  'LOR': np.random.uniform(1.0, 5.0, size=100),
  'CGPA': np.random.uniform(6.0, 10.0, size=100),
  'Research': np.random.randint(0, 2, size=100),
  'Chance of Admit': np.random.uniform(0.0, 1.0, size=100)
})
# Define the categories and colors
category = ['GRE Score', 'TOEFL Score', 'University Rating', 'SOP', 'LOR', 'CGPA',
'Research', 'Chance of Admit']
color = ['yellowgreen', 'gold', 'lightskyblue', 'pink', 'red', 'purple', 'orange', 'gray']
# Create the subplots
fig, axs = plt.subplots(nrows=4, ncols=2, figsize=(14, 8))
for i in range(8):
  row = i // 2
  col = i \% 2
  axs[row, col].hist(data[category[i]], color=color[i], bins=10)
```

axs[row, col].set_title(category[i])

plt.show()

plt.subplots_adjust(hspace=0.7, wspace=0.2)

```
# In[33]:
from sklearn.preprocessing import MinMaxScaler
sc = MinMaxScaler()
x=sc.fit transform(x)
X
# In[15]:
x=data.iloc[:,0:7].values
X
# In[16]:
y=data.iloc[:,7:].values
y
# In[15]:
from sklearn.model_selection import train_test_split
x = [[0],[1],[2],[3]]
y = [0, 1, 2, 3]
x_train, x_test, y_train, y_test = train_test_split(x,y, test_size=0.30,random_state=101)
# In[18]:
y_train=(y_train>0.5)
y_train
# In[34]:
y_test=(y_test>0.5)
# In[]:
#train_data = pd.read_csv(r"C:\Users\vigne\Intelligent Admissions The Future of University
Decision Making with Machine Learning\Dataset\Admission_Predict.csv")
#x_train = train_data
#y_train = train_data
# In[1]:
```

```
from sklearn.linear_model.logistic import LogisticRegression cls
=LogisticRegression(random_state = 2)
lr=cls.fit(x_train,y_train)
y_pred = lr.predict(x_test)
print(y_pred)
# In[2]:
y_pred =lr.predict(x_test)
print(y_pred)
# In[3]:
#libraries to train neural network
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.layers import Dense, Activation, Dropout
from tensorflow.keras.optimizers import Adam
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras import layers
# Define the model
model = keras.Sequential([
  layers.Dense(64, activation='relu'),
  layers.Dense(10)
1)
# Compile the model
model.compile(optimizer=tf.keras.optimizers.Adam(0.01),
        loss=tf.keras.losses.CategoricalCrossentropy(from_logits=True),
        metrics=['accuracy'])
# Train the model
model.fit(x_train, y_train, epochs=10, validation_data=(x_val, y_val))
```

```
# In[4]:
model=keras.Sequential()
model.add(Dense(7,activation ='relu',input_dim=7))
model.add(Dense(7,activation='relu'))
model.add(Dense(1,activation='linear'))
model.summary()
# In[]:
#from sklearn.linear_model import LinearRegression
#model = LinearRegression()
#model.fit(x_train, y_train)
model.fit(x_train, y_train, batch_size = 20, epochs = 100)
# In[44]:
model.compile(loss = 'binary_crossentropy', optimizer = 'adam',metrics = ['accuracy'])
# In[5]:
model.fit(x_train, y_train, batch_size = 20, epochs = 100)
# In[6]:
from sklearn.metrics import accuracy_score
train_prediction = model.predict(x_train)
print(train_predictions)
# In[10]:
print(classification report(y test.pred))
# In[8]:
train_acc = model.evaluate(x_train, y_train, verbose=0)[1]
print(train_acc)
# In[9]
train_acc = model.evaluate(x_test, y_test, verbose=0)[1]
```

```
print(train_acc
# In[11]:
pred=model.predict(x_test)
pred = (pred > 0.5)
pred
# In[12]:
from sklearn.metrics import accuracy_score,recall_score,roc_auc_score,confusion_matrix
print("\nAccuracy score: %f" %(accuracy_score()*100))
print("Recall score : %f" %(recall_score(y_test,y_pred)*100))
print("ROC score : %f\n" %(roc_auc_score(y_test,y_pred)*100))
print(confusion_matrix(y_test,y_pred))
# In[13]:
#ANN Model
from sklearn.metrics import accuracy_score,recall_score,roc_auc_score,confusion_matr:
print(classification_report(y_train,pred))
# In[14]:
from sklearn.metrics import accuracy_score,recall_score,roc_auc_score
print(classification_report(y_test,pred))
# In[15]:
model.save('model.h5')
```