Diabetes Prediction Model



TEAM: Data_wizards

Anthony Ngatia

Naomi Rotich

Elsie Nduta

Mitchele Okubasu

Content Outline

Topics for discussion

01 Project Overview

02 Project Roadmap

03 Introduction

04 Visualizations

05 Modelling

Key Challenges

Conclusion & Recomendation



Project Overview

- Dataset: Extensive medical records of patients, including demographics, clinical measurements, and medical history.
- Methodology: CRISP-DM framework Business Understanding, Data
 Understanding, Data Preparation,
 Modeling, Evaluation and Deployment.
- Tools: Google Colab, Jupyter Notebook, Github, Spyder(Anaconda)etc.



Project Roadmap

CRISP-DM Process



Business Understanding

Exploratory Data Analysis



Deployment Analysis



Modelling & Evaluation

Data Preparation

Data

Understanding





Introduction

Objective:

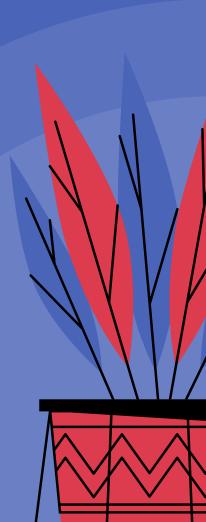
To develop an accurate and reliable predictive model for early detection of diabetes.

Why it matters:

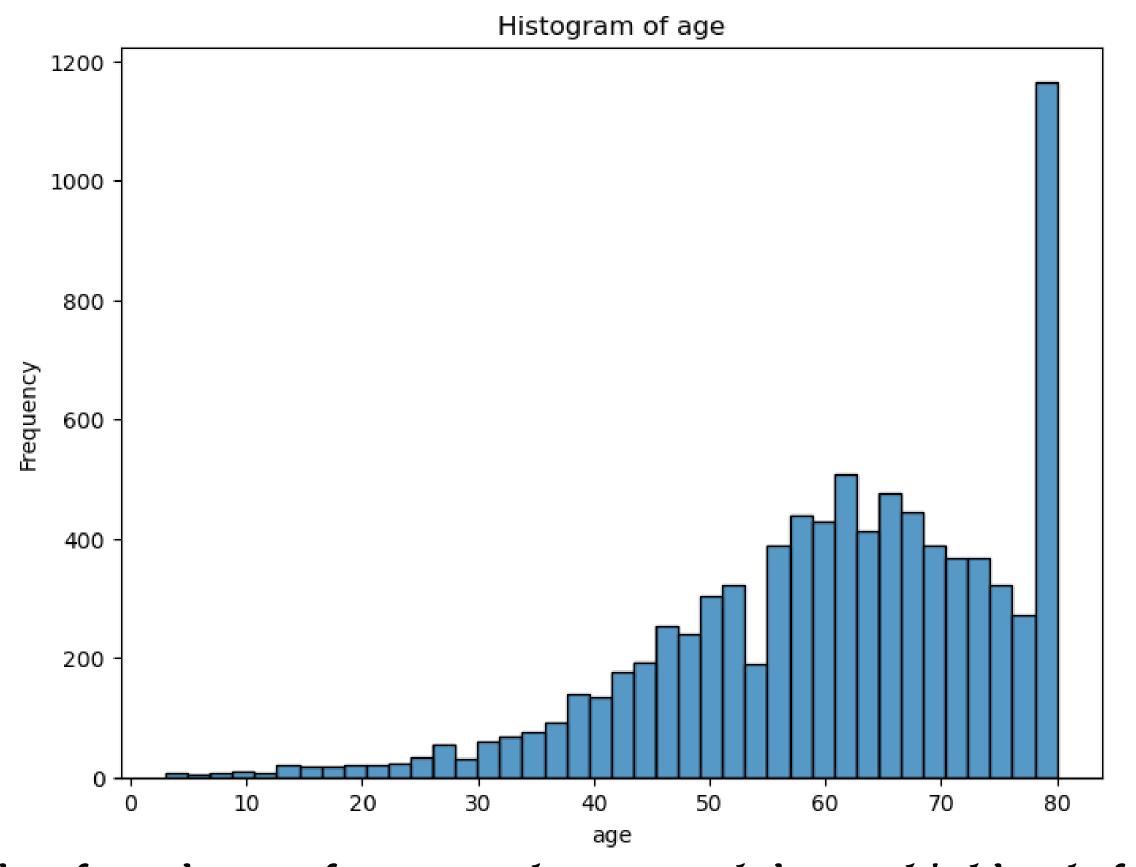
Early detection can lead to timely intervention and better patient outcomes.

• Data-driven approach:

Utilizing advanced data analysis and machine learning techniques.



From what age does Diabetes occurence increase?



It appears that from the age of 40 onwards most people have a likelihood of having diabetes

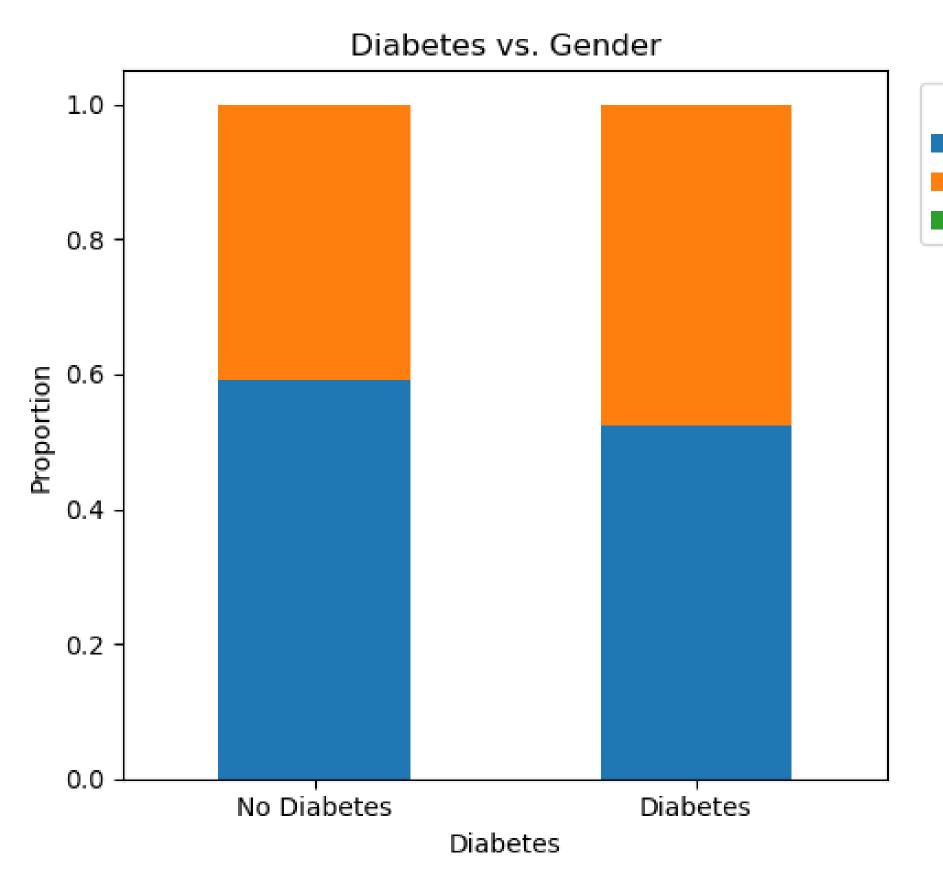
What Gender has more occurrences of Diabetes?

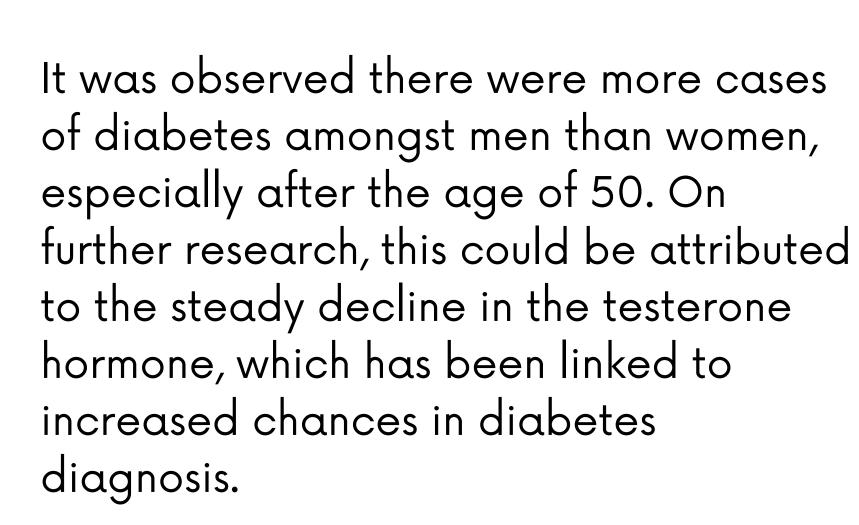
Gender

Female

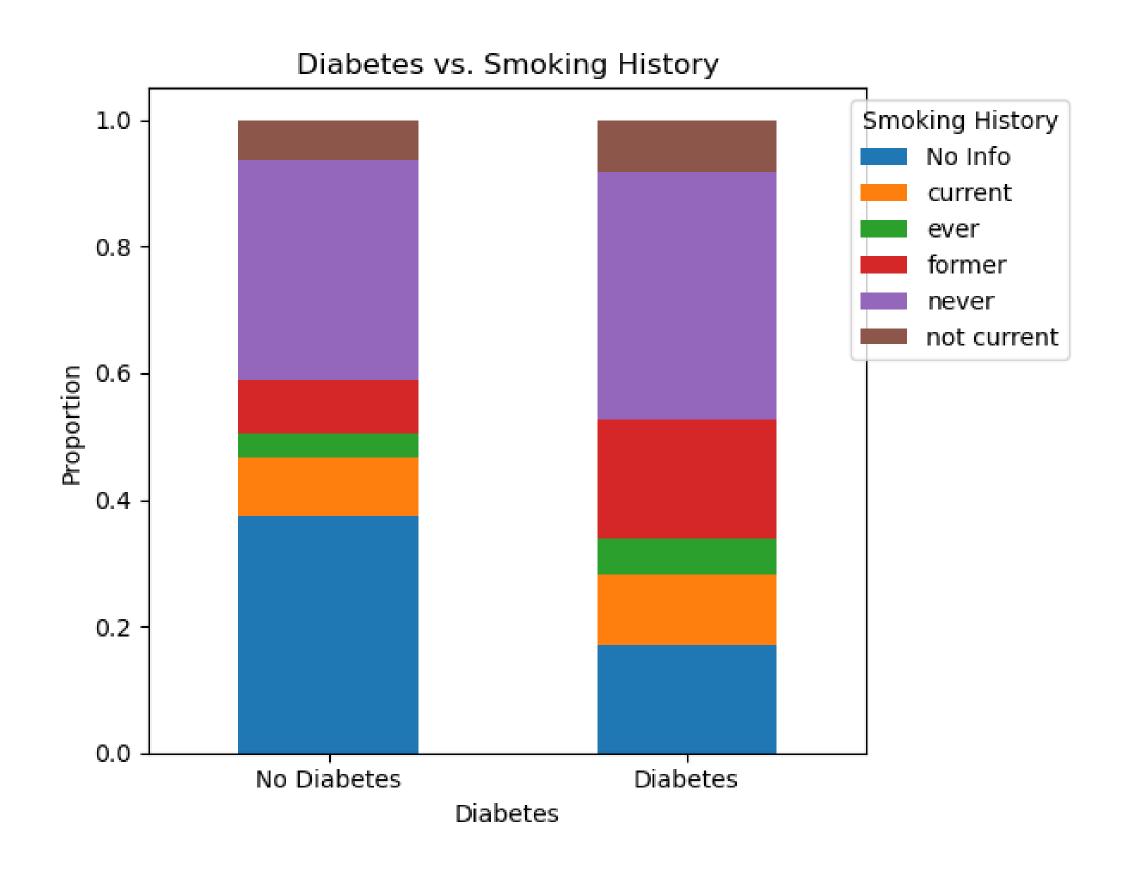
Male

Other

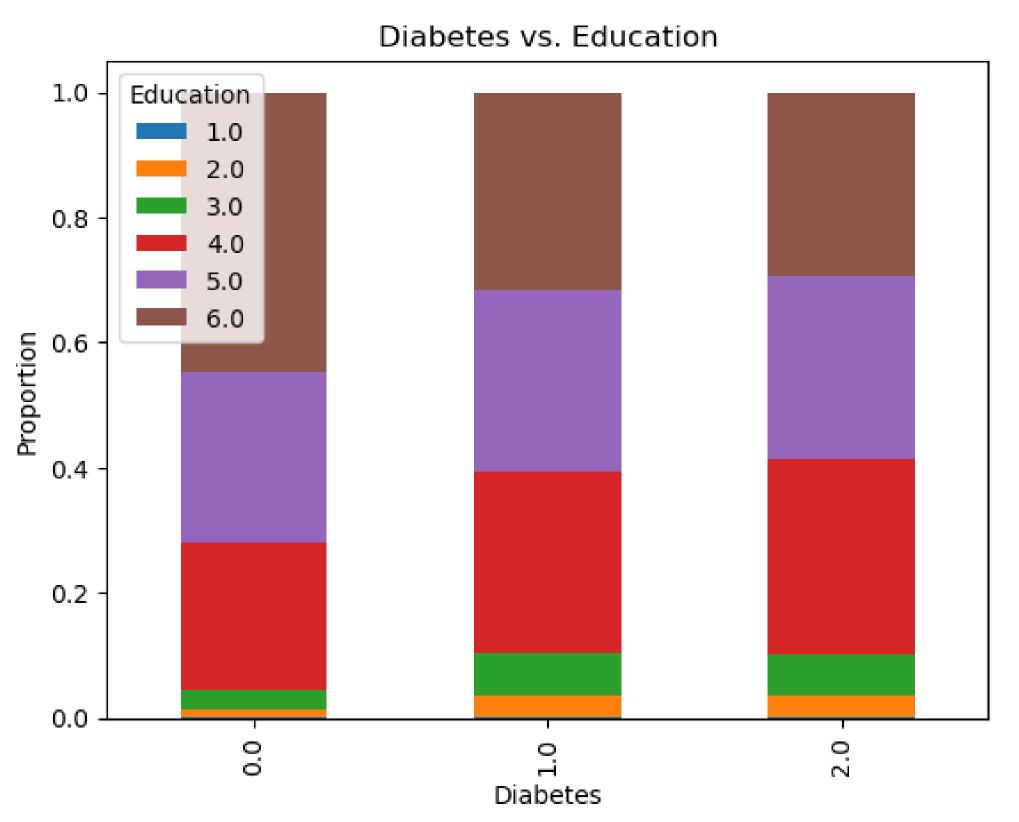




What's the smoking comparison between those with diabetes and no_Diabetes?

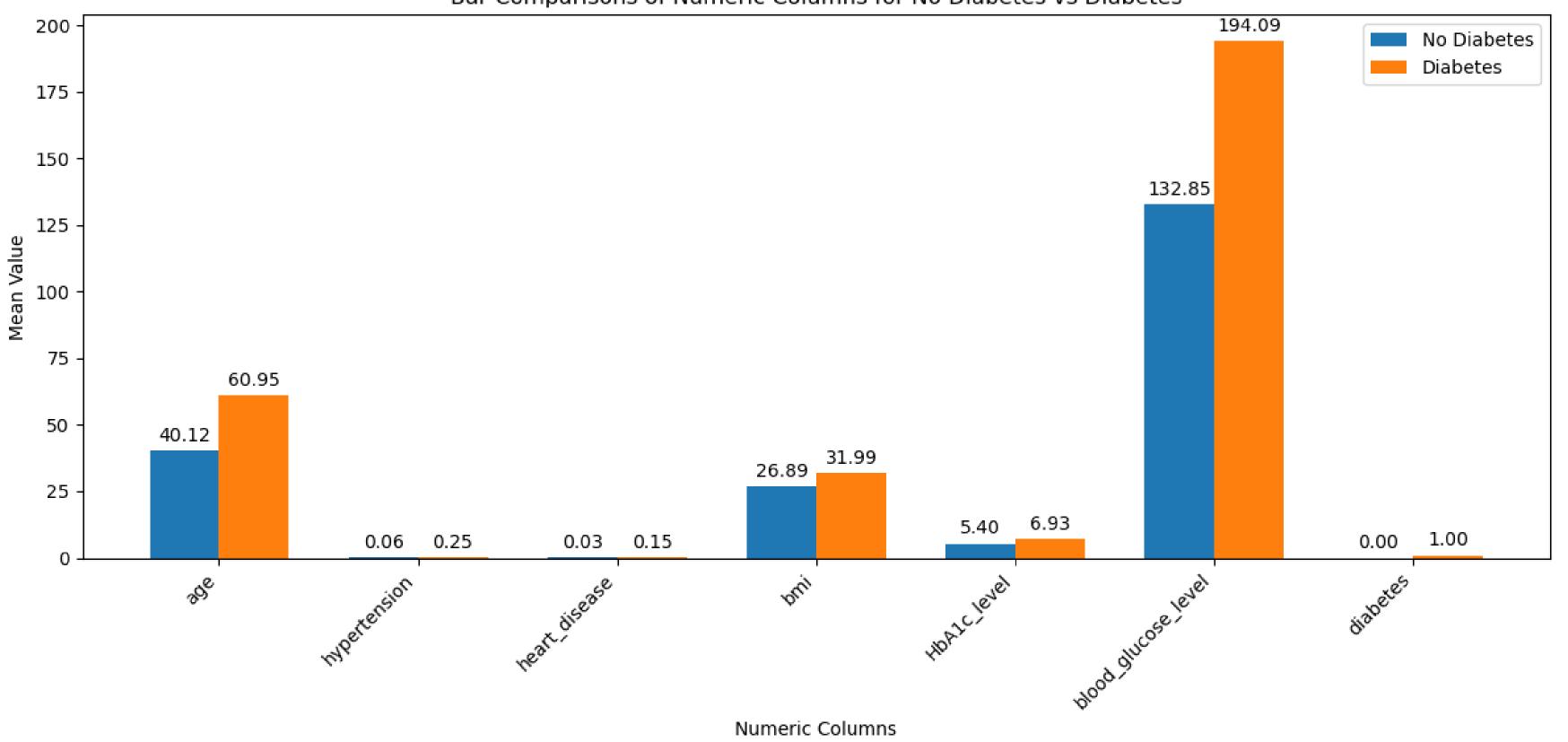


What's the comparison of education levels and those with diabetes and no_Diabetes?



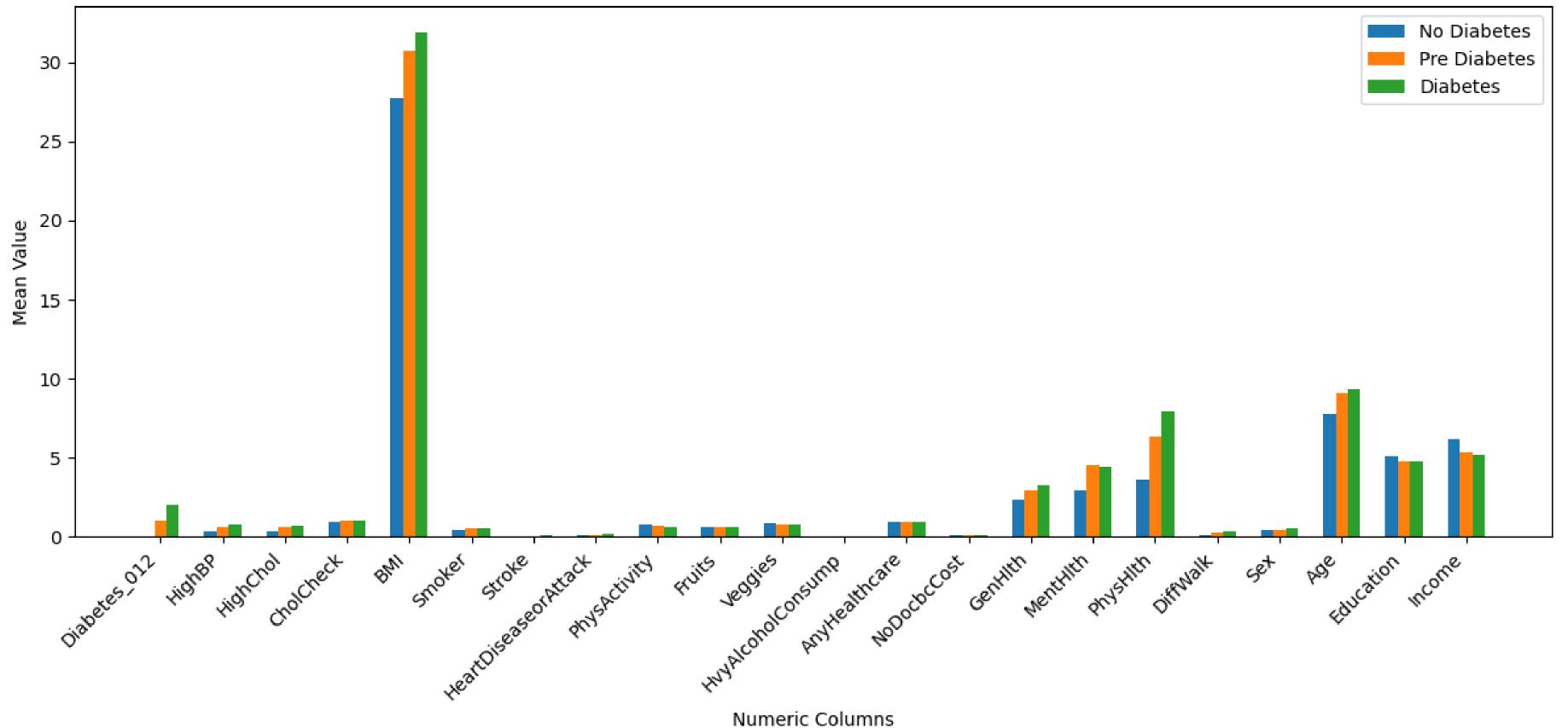
Comparisons between those with diabetes and no_Diabetes?

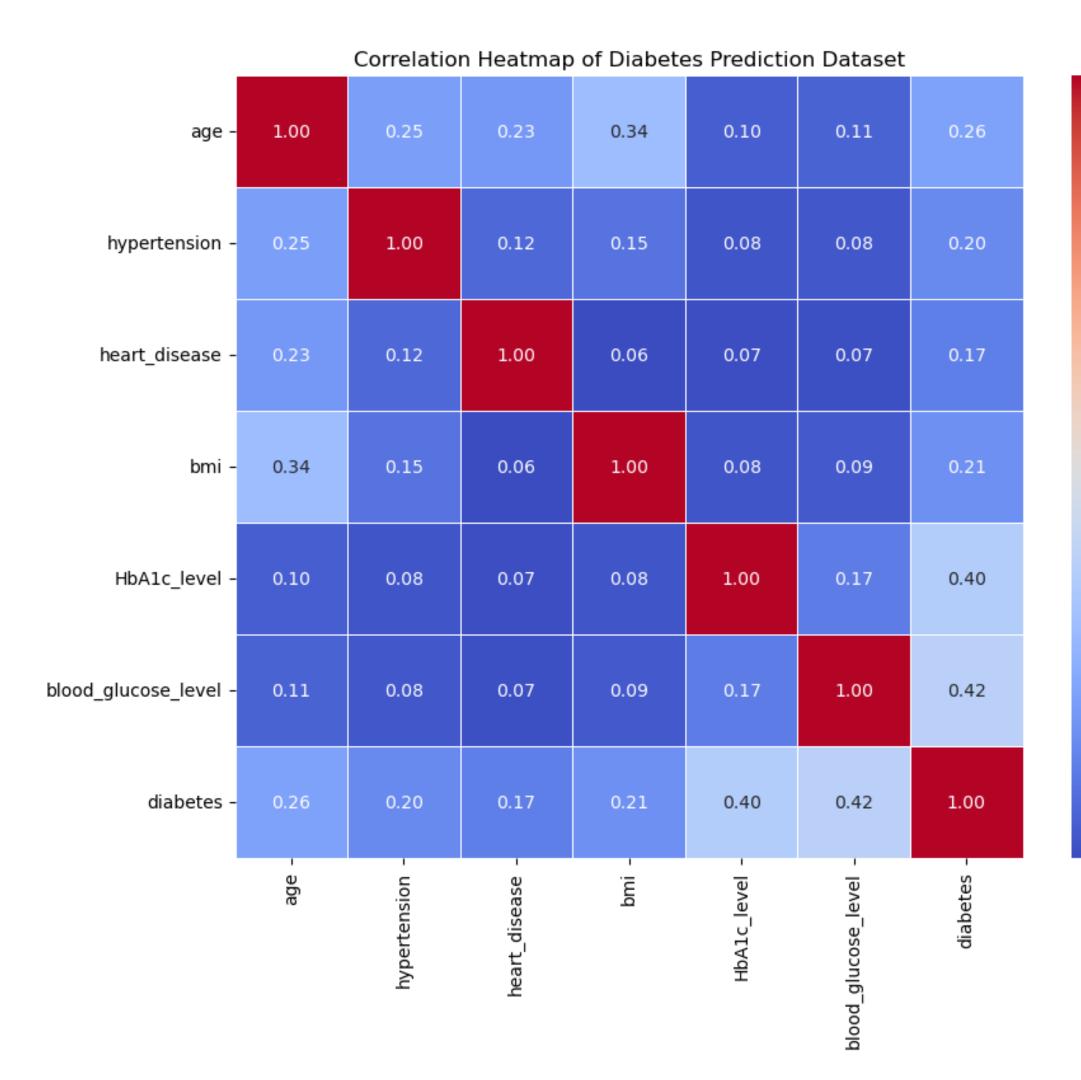




Comparisons between those with diabetes and no_Diabetes?

Bar Comparisons of Numeric Columns for Pre Diabetes vs Diabetes





Correlation of all the factors

- 0.8

- 0.4

- 0.2

Modelling

XGBOOST

CNN MODEL

SVM MODEL

Decision Tree



Model Evaluation



XGBOOST

Precision:

0.99

Recall:

0.84

F1 Score:

0.84

CNN

Precision:

0.98

Recall:

0.67

F1 Score:

0.80

SVM MODEL

Precision:

0.89

Recall:

0.89

F1 Score:

0.89

Decision Tree

Precision:

0.96

Recall:

0.96

F1 Score:

0.96

Function that prompts user to input gender, age,...and returns a prediction based on the input

model: **Decision Tree**

```
predict_diabetes(model)
Enter gender (Male/Female): Female
Enter age: 60
Enter hypertension (0 for No, 1 for Yes): 0
Enter heart disease (0 for No, 1 for Yes): 0
Enter smoking history (never/No Info/current/former/ever/not current): former
Enter RMT: 9
Enter HbA1c level: 6.5
Enter blood glucose level: 200
Based on the input, the person is predicted to NOT have diabetes.
```

```
predict_diabetes(model)
Enter gender (Male/Female): Female
Enter age: 60
Enter hypertension (0 for No, 1 for Yes): 1
Enter heart disease (0 for No, 1 for Yes): 0
Enter smoking history (never/No Info/current/former/ever/not current): ever
Enter BMI: 27
Enter HbA1c level: 7
Enter blood glucose level: 150
Based on the input, the person is predicted to have diabetes.
```

Key Challenges

Imbalanced dataset

Model generalization to diverse patient populations

CONCLUSIONS

The gender with more occurences of diabetes are Females.

Most smokers seem to quit smoking after being diagnosed with diabetes this is because smoking habits have a probability of causes smoking is a significant risk factor for developing type 2 diabetes.

As income increases there are lesser people with diabetes these could be correlated with their ability to lead healthier lifestlyes .

Those with higher level of education also have lesser occurences of diabetes this could be associated by their ability to do better research or an understnding of the importance to led healthier lifestyles.

RECOMMENDATIONS



Collection of a more balanced data(no diabetes and diabetes) would produce a better performing model that wont be biased after training.