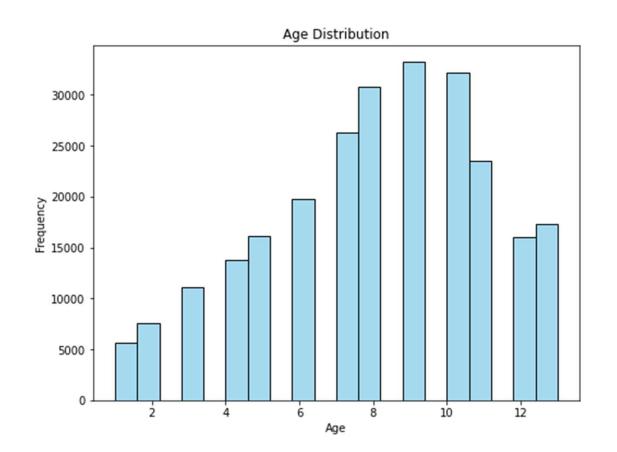


## **Business and Data Understanding**

#### **Problem Statement**

- In the U.S., 1 in 3 adults are prediabetic, yet 80% remain unaware of their condition. Without timely intervention, most will develop type 2 diabetes within 5 years, leading to severe health complications and significant financial burdens—\$413 billion was spent on diabetes care in 2022 alone.
- To address this challenge, Hanover is launching the Diabetes Education and Engagement Program (DEEP). This initiative leverages data analysis to identify at-risk individuals for early intervention, improve detection rates, and empowers the community through education to prevent and manage diabetes effectively

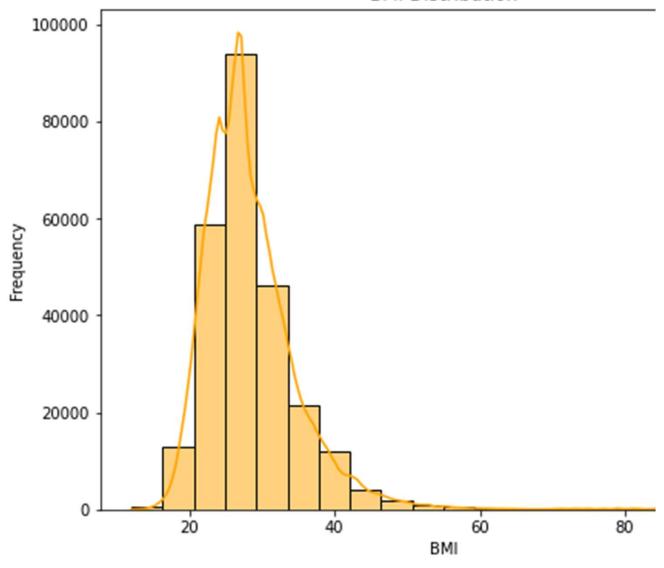


# **Business and Data Understanding**

### **Demographic Distribution**

- Age ranges from 18-80 years, peak 55-69 years old.
- Sex: women outnumber men 56% to 44%
- Income: The majority income above \$35000

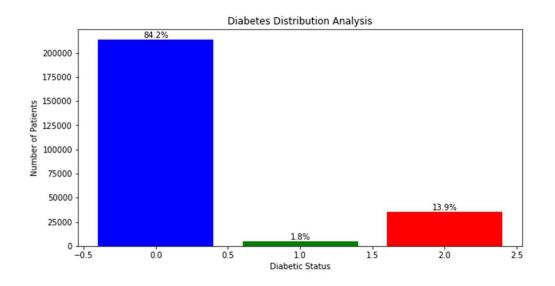
### **BMI Distribution**



# **Business and Data Understanding**

- Health indicators
- Most BMI falls between 20-40.
- More Non-Smokers than Smokers in the dataset (140000: 105000)
- Physical Activities: majority of individuals are "Active" - females are more physically active overall.

### **Business and Data Understanding**



#### **Diabetes Distribution**

- **21,3703** labeled"0" (nondiabetic)
- 4631 labeled as "1" (pre-diabetes)
  1.82%
- **35,346** labeled as "2" ( diabetes) **13.9**%

# Classification System: Technique and Model selection

- XGBoost creates a predictive machine learning model of the data
- SHAP (SHapley Additive exPlanations)

#### **Evaluation Metrics**

- Accuracy, precision, recall, and F1-score for balanced and relevant predictions.
- SHAP fidelity evaluation



# Classification System: Technique and Model selection

### **Target Variable Description**

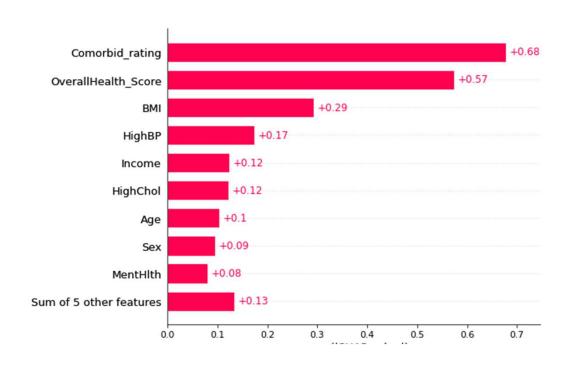
Diabetes & Pre-Diabetes = **Diabetes\_status** 

**Non-Diabetes** as the second category.

Early Detection Focus: flags anyone at risk of diabetes,.

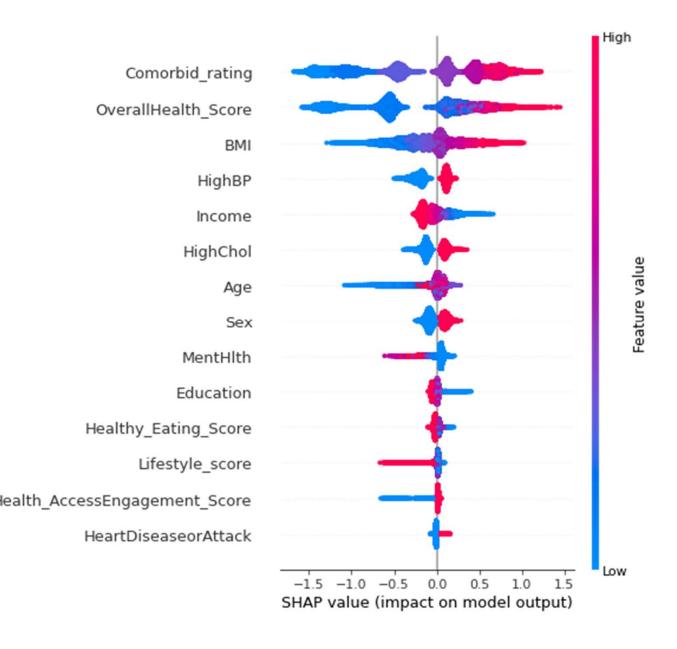
Preventive Action: Treats both diabetic and pre-diabetic as high-risk groups, prompting similar preventive measures.

### Analysis of the feature importances



### • Feature Importance Order:

- Top contributors to the model's predictions
  - Comorbid\_rating
  - OverallHealth\_Score
  - BMI

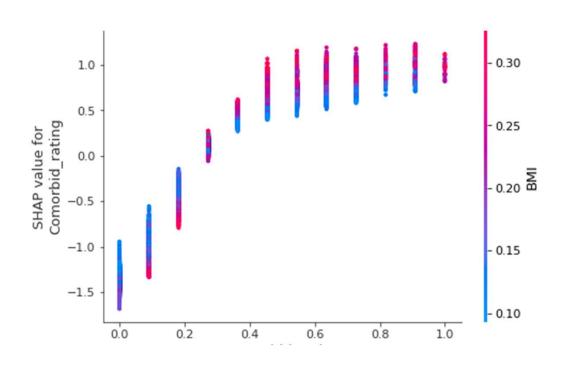


# Analysis of the feature importances

## Features impact on model's output

- High importance features
- Moderate importance features
- Low importance features

### **Analysis of the feature importances**



- Feature Interactions:
- High Comorbid\_rating & low interacting feature (blue) moderate SHAP values, (protective effect) offsets the severity of comorbidities.
- High Comorbid\_rating & high interacting feature (pink) strongest SHAP values, maximum contribution to risk.

# Recommended Interventions

Target

Target High-Risk Groups: Individuals with high Comorbid\_rating, poor health, low income, and unhealthy lifestyles.

Leverage

Leverage Protective Factors: Promote healthy eating, higher education, better healthcare access, and income growth.

Address

Address Interactions: Target scenarios where multiple high-risk features compound effects to implement tailored





Improve model predictive performance and accuracy

- Validate Findings
- Feature Engineering
- Enhance Model Interpretability
- Cluster-Based Analysis

### **Questions**



Thank You!

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