

## STEPS



Problem Definition

2

**Data Collection** 

3

Exploratory
Data Analysis
and Data
Visualization



Model



Model Testing and Results



# GANTT

		MONTHS							
No	Work Packages	1	2	3	4	5	6	7	8
1	Problem Detection								
2	Dataset Collection								
3	EDA and Data Visualizations								
4	Model Creation								
5	Model Testing and Interpretation								





# 1.PROBLEM DEFINITON

Obesity is a serious health problem linked to heart diseases, diabetes, and chronic illnesses, reducing quality of life. Data science methods help analyze the factors causing obesity and create personalized health solutions.



### 2.DATA COLLECTION

A suitable dataset has been found on Kaggle for conducting a data science project related to the topic. You can access the dataset through the following link: kaggle.com/datasets/lesumitkumarroy/obe sity-data-set



#### 3. EXPLORATORY DATA ANALYSIS AND DATA VISUALIZATION

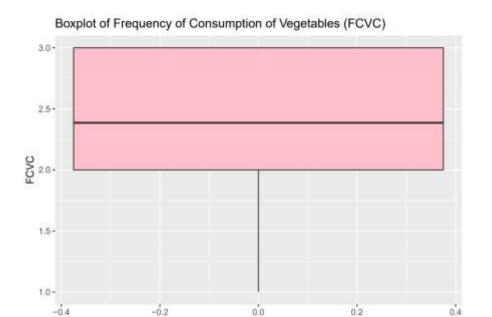
#### **DEFINITION OF VARIABLES**

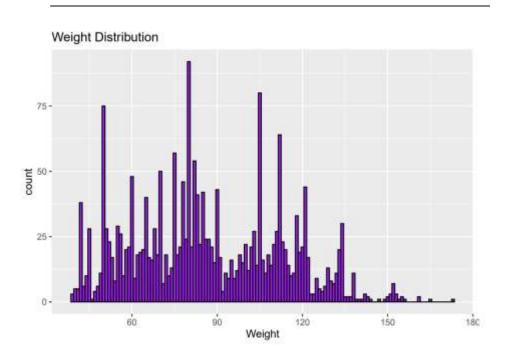
- **Gender**: Gender (Male/Female).
- Age: Age (numeric value).
- **Height**: Height (in meters).
- Weight: Weight (in kilograms).
- family\_history\_with\_overweight: Family history of obesity (Yes/No).
- **FAVC**: Habit of consuming high-calorie food (Yes/No).
- **FCVC**: Frequency of vegetable consumption (a value between 1-3).
- **NCP**: Number of main meals per day (numeric value).

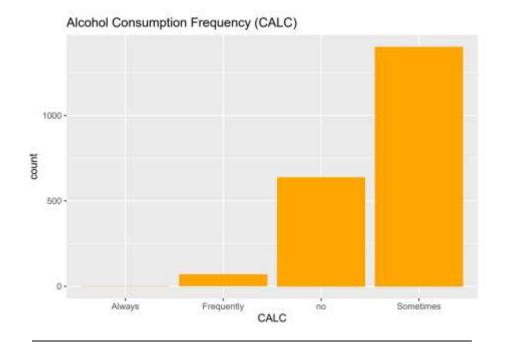
- **SMOKE**: Smoking habit (Yes/No).
- **CH2O**: Daily water consumption (in liters).
- **SCC**: Habit of controlling calories (Yes/No).
- **FAF**: Weekly physical activity time (in hours).
- **TUE**: Daily technology usage time (in hours).
- CALC: Alcohol consumption frequency (None/Weekly/Daily/Special Occasions).
- MTRANS: Mode of daily transportation (Car/Motorbike/Bicycle/Walking/Public transport).
- NObeyesdad: Target variable. Obesity level (Normal weight, Overweight, Slightly obese, Severely obese, etc.).

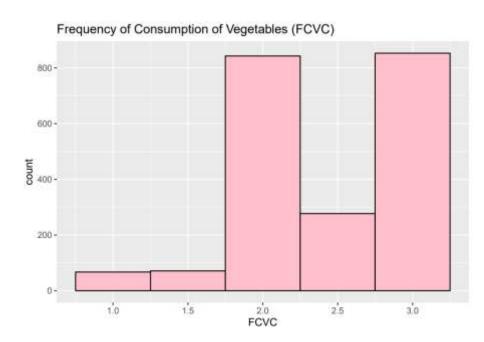
```
# Check the number of rows and columns
str(data)
## spc_tbl_ [2,111 x 17] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
## $ Gender
                                  : chr [1:2111] "Female" "Female" "Male" "Male" ...
## $ Age
                                  : num [1:2111] 21 21 23 27 22 29 23 22 24 22 ...
## $ Height
                                  : num [1:2111] 1.62 1.52 1.8 1.8 1.78 1.62 1.5 1.64 1.78 1.72 ...
## $ Weight
                                  : num [1:2111] 64 56 77 87 89.8 53 55 53 64 68 ...
## $ family_history_with_overweight: chr [1:2111] "yes" "yes" "yes" "no" ...
## $ FAVC
                                  : chr [1:2111] "no" "no" "no" "no" ...
## $ FCVC
                                  : num [1:2111] 2 3 2 3 2 2 3 2 3 2 ...
## $ NCP
                                  : num [1:2111] 3 3 3 3 1 3 3 3 3 3 ...
## $ CAEC
                                  : chr [1:2111] "Sometimes" "Sometimes" "Sometimes" ...
## $ SMOKE
                                  : chr [1:2111] "no" "yes" "no" "no" ...
## $ CH20
                                  : num [1:2111] 2 3 2 2 2 2 2 2 2 2 ...
## $ SCC
                                  : chr [1:2111] "no" "yes" "no" "no" ...
## $ FAF
                                  : num [1:2111] 0 3 2 2 0 0 1 3 1 1 ...
## $ TUE
                                  : num [1:2111] 1 0 1 0 0 0 0 0 1 1 ...
                                  : chr [1:2111] "no" "Sometimes" "Frequently" "Frequently" ...
## $ CALC
                                  : chr [1:2111] "Public Transportation" "Public Transportation" "Pub
## $ MTRANS
                                  : chr [1:2111] "Normal_Weight" "Normal_Weight" "Ove
## $ NObeyesdad
## - attr(*, "spec")=
    .. cols(
         Gender = col_character(),
    .. Age = col_double(),
    .. Height = col double(),
    .. Weight = col_double(),
    .. family_history_with_overweight = col_character(),
    .. FAVC = col_character(),
    .. FCVC = col_double(),
    .. NCP = col double(),
         CAEC = col_character(),
         SMOKE = col character(),
         CH20 = col_double(),
         SCC = col_character(),
         FAF = col double(),
    .. TUE = col_double(),
         CALC = col_character(),
    .. MTRANS = col_character(),
         NObeyesdad = col_character()
     . .
```

- attr(\*, "problems")=<externalptr>

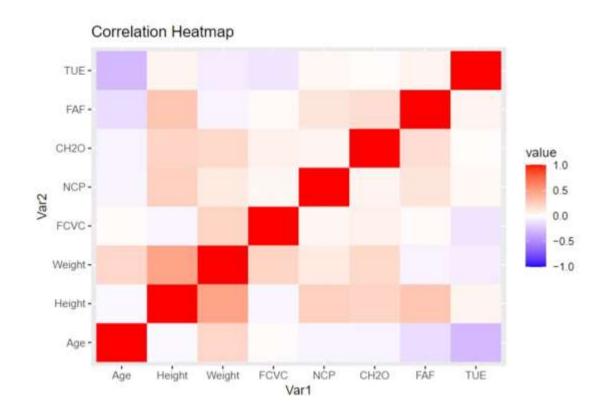












As a result of examining and visualizing the data, the variables that are considered to directly affect obesity and are relatively easier to address with solutions have been identified as follows;

- FCVC (Frequency of Vegetable Consumption): Increase vegetable consumption.
- FAVC (Consumption of High-Calorie Food): Reduce the consumption of high-calorie foods.
- FAF (Physical Activity Time): Increase physical activity.
- CH2O (Water Consumption): Increase daily water consumption.
- TUE (Technology Usage Time): Reduce technology usage time to create more time for physical activity.

In this context, we can begin exploring the obesity situation by focusing on the positive changes in these variables.



### 4.MODEL

Obesity status (NObeyesdad) is usually determined using a metric like Body Mass Index (BMI). BMI is a parameter that helps estimate a person's body fat percentage based on their height and weight.

BMI is calculated using the following formula:

```
BMI=Weight (kg)/Height (m)2
```

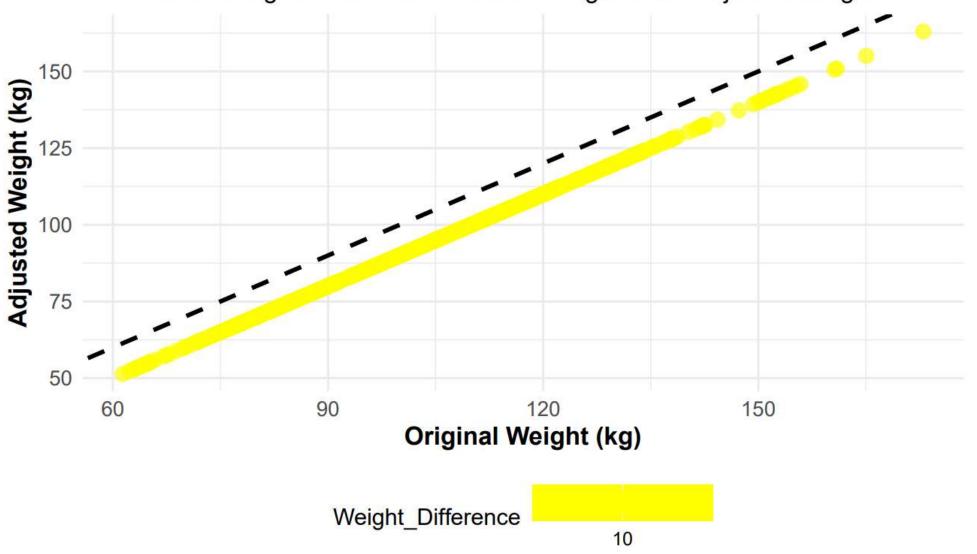




```
# New data for prediction based on mean values with specific adjustments
new_data <- data.frame(</pre>
 FCVC = mean(overweight_obese_data$FCVC) * 1.02,
  # 2% increase in vegetable consumption
  FAVC = mean(overweight_obese_data$FAVC) * 0.98,
  # 2% decrease in high-calorie food consumption
  FAF = mean(overweight_obese_data$FAF) * 1.05,
  # 5% increase in physical activity
  CH20 = mean(overweight_obese_data$CH20) * 1.02,
  # 2% increase in water consumption
  TUE = mean(overweight_obese_data$TUE) * 0.95
  # 5% decrease in technology use
```

#### **Original vs Adjusted Weights**

Visualizing the Difference Between Original and Adjusted Weights





```
##
 ## Call:
## lm(formula = Weight ~ FCVC + FAVC + FAF + CH20 + TUE, data = train_data)
## Residuals:
       Min
               10 Median
                                     Max
## -41.990 -11.569 -1.485 13.150 53.117
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                          5.0468 1.333
                6.7254
                                           0.183
## FCVC
               14.0693
                         1.0108 13.919 < 2e-16 ***
## FAVC
               25.6381
                          2.0906 12.263 < 2e-16 ***
## FAF
                3.4866
                         0.6673 5.225 2.14e-07 ***
                4.7735
                          0.8750 5.455 6.22e-08 ***
 ## CH20
 ## TUE
               -0.2230
                          0.9223 -0.242
                                         0.809
 ## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
 ##
## Residual standard error: 16.09 on 964 degrees of freedom
## Multiple R-squared: 0.3049, Adjusted R-squared: 0.3013
## F-statistic: 84.55 on 5 and 964 DF, p-value: < 2.2e-16
# Limit the weight change predictions between -10 and 10
predicted_weight_changes_scaled <-
  pmin(pmax(predicted_weight_changes_scaled, -10), 10)
```

# 5. MODEL TESTING AND RESULTS







# THANK YOU FOR LISTENING!