# Project 3

This is the dataset you will be working with:

```
food <- readr::read_csv("https://wilkelab.org/DSC385/datasets/food_coded.csv")
food</pre>
```

```
## # A tibble: 125 x 61
##
      GPA
            Gender breakfast calories chicken calories day calories scone coffee
##
      <chr>>
             <dbl>
                        <dbl>
                                          <dbl>
                                                        <dbl>
                                                                       <dbl>
                                                                               <dbl>
##
   1 2.4
                  2
                            1
                                            430
                                                          NaN
                                                                          315
                                                                                   1
    2 3.654
                            1
                                                            3
                                                                                   2
##
                  1
                                            610
                                                                          420
##
   3 3.3
                  1
                            1
                                            720
                                                            4
                                                                          420
                                                                                   2
##
   4 3.2
                  1
                            1
                                            430
                                                            3
                                                                          420
                                                                                   2
   5 3.5
##
                  1
                            1
                                            720
                                                            2
                                                                          420
                                                                                   2
                            1
                                                            3
##
   6 2.25
                  1
                                            610
                                                                          980
                                                                                   2
##
   7 3.8
                  2
                            1
                                            610
                                                            3
                                                                          420
                                                                                   2
   8 3.3
                  1
                            1
                                                            3
                                                                          420
                                                                                   1
##
                                            720
                            1
   9 3.3
                  1
                                            430
                                                                                   1
##
                                                          NaN
                                                                          420
                            1
                                                                                   2
## 10 3.3
                  1
                                            430
                                                                          315
## # ... with 115 more rows, and 54 more variables: comfort_food <chr>,
       comfort food reasons <chr>, comfort food reasons coded...10 <dbl>,
## #
       cook <dbl>, comfort food reasons coded...12 <dbl>, cuisine <dbl>,
## #
## #
       diet current <chr>, diet current coded <dbl>, drink <dbl>,
       eating_changes <chr>, eating_changes_coded <dbl>,
## #
       eating changes coded1 <dbl>, eating out <dbl>, employment <dbl>,
## #
       ethnic food <dbl>, exercise <dbl>, father education <dbl>, ...
## #
```

A detailed data dictionary for this dataset is available here.

(https://wilkelab.org/DSC385/datasets/food\_codebook.pdf) The dataset was originally downloaded from Kaggle, and you can find additional information about the dataset here. (https://www.kaggle.com/borapajo/food-choices/version/5)

**Question:** Is GPA related to student income, the father's educational level, or the student's perception of what an ideal diet is?

To answer this question, first prepare a cleaned dataset that contains only the four relevant data columns, properly cleaned so that numerical values are stored as numbers and categorical values are represented by humanly readable words or phrases. For categorical variables with an inherent order, make sure the levels are in the correct order.

In your introduction, carefully describe each of the four relevant data columns. In your analysis, provide a summary of each of the four columns, using summary() for numerical variables and table() for categorical variables.

Then, make one visualization each for student income, father's educational level, and ideal diet, and answer the question separately for each visualization. The three visualizations can be of the same type.

## Hints:

1. Use case when() to recode categorical variables.

- 2. Use fct relevel() to arrange categorical variables in the right order.
- 3. Use as.numeric() to convert character strings into numerical values. It is fine to ignore warnings about NA s introduced by coercion.
- 4. Nan stands for Not a Number and can be treated like NA. You do not need to replace Nan with NA.
- 5. When using table(), provide the argument useNA = "ifany" to make sure missing values are counted: table(..., useNA = "ifany").

Introduction: The dataset used in the study is the <code>food\_coded</code> dataset provided by <code>Kaggle</code>. The data consists of survey answers from the students of <code>Mercyhurst University</code> on various preferences of food as well as some other general information on the student such as <code>GPA</code> and <code>weight</code>. For this analysis, we will use the <code>income</code>, <code>father\_education</code>, and <code>ideal\_diet\_coded</code> data columns to predict the students <code>GPA</code>. The <code>GPA</code> is the standard four point system used at most universities. The <code>income</code> data is coded from 1 to 6 representing ranges of income. The <code>father\_education</code> data is coded from 1 to 5 indicating what level of education the student's father has earned such as high school diploma or college degree. The <code>ideal\_diet\_coded</code> data is coded from 1 to 8 with various answers from students on what they consider to be important to an ideal diet in the context of their current diet. Answers to this survey range from portion control to adding veggies. The data is cleaned, isolated, and displayed below.

```
food <- food %>%
  mutate(GPA_numeric = as.numeric(GPA),
         income_readable = case_when(
           income == 1 ~ "less than $15,000",
           income == 2 \sim "$15,001 to $30,000",
           income == 3 ~ "$30,001 to $50,000",
           income == 4 \sim "$50,001 \text{ to } $70,000",
           income == 5 \sim "$70,001 to $100,000"
           income == 6 \sim "higher than $100,000",
           TRUE ~ NA character
           ),
         father education readable = case when(
           father education == 1 ~ "less than high school",
           father education == 2 ~ "high school degree",
           father_education == 3 ~ "some college degree",
           father education == 4 ~ "college degree",
           father_education == 5 ~ "graduate degree",
           TRUE ~ NA_character_
           ),
         ideal diet readable = case when(
           ideal_diet_coded == 1 ~ "portion control",
           ideal_diet_coded == 2 ~ "eating healthier food",
           ideal_diet_coded == 3 ~ "balance",
           ideal diet coded == 4 ~ "less sugar",
           ideal_diet_coded == 5 ~ "home cooked/organic",
           ideal diet coded == 6 ~ "current diet",
           ideal diet coded == 7 ~ "more protein",
           ideal diet_coded == 8 ~ "unclear",
           TRUE ~ NA character
           )
         )
```

```
## Warning in mask$eval_all_mutate(quo): NAs introduced by coercion
summary(food$GPA_numeric)
##
      Min. 1st Qu. Median
                                                        NA's
                              Mean 3rd Qu.
                                               Max.
##
     2.200
             3.200
                     3.500
                              3.416
                                      3.700
                                              4.000
                                                           5
as.data.frame(table(food$income readable, useNA = "ifany"))
##
                     Var1 Freq
## 1
       $15,001 to $30,000
                             7
## 2
       $30,001 to $50,000
                             17
       $50,001 to $70,000
## 3
                             20
## 4
     $70,001 to $100,000
                             33
## 5 higher than $100,000
                             41
## 6
        less than $15,000
                              6
## 7
                      <NA>
                              1
as.data.frame(table(food$father_education_readable, useNA = "ifany"))
##
                      Var1 Freq
## 1
            college degree
                              46
           graduate degree
## 2
                              28
## 3
        high school degree
                              34
## 4 less than high school
                              4
## 5
       some college degree
                              12
## 6
                       <NA>
                               1
as.data.frame(table(food$ideal diet readable, useNA = "ifany"))
##
                      Var1 Freq
## 1
                   balance
                              17
              current diet
## 2
                              13
## 3 eating healthier food
                              44
## 4
       home cooked/organic
                              15
## 5
                less sugar
                              6
## 6
              more protein
                              16
## 7
           portion control
                              11
## 8
                   unclear
```

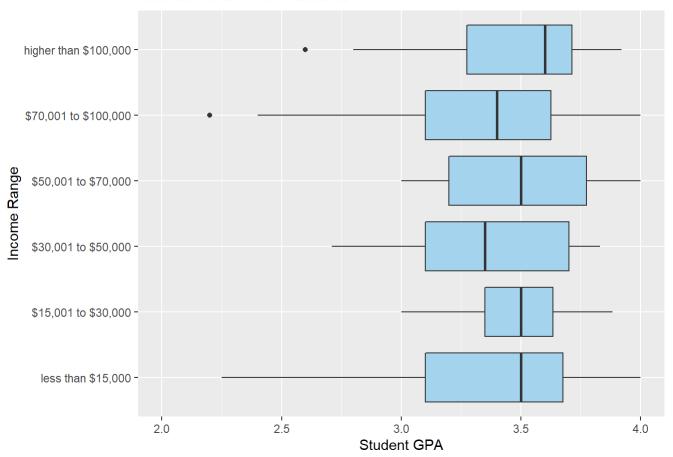
**Approach:** Because the independent data given is in all cases categorical, we will use a series of boxplots. The boxplots will be separated by the independent variable and aligned on the same axis to make clear comparisons of the student's GPA. Boxplots are also beneficial in that they show the medians of the student's GPAs for each category; this allows us to track a trend if one can be seen. The GPA axis of the plots will limit from 2 to 4 to keep them alike and not skew the scale across each. It will also give context with relation to the bottom and top performers.

# Analysis:

Here we filter out all GPAs that are invalid ( N/A ) and we filter out all income values that are invalid ( N/A ). We plot the boxplots in order of the income labels for ease of readability.

```
food %>%
  filter(
  !is.na(GPA_numeric),
  !is.na(income_readable)
  ) %>%
  ggplot(aes(GPA_numeric, fct_reorder(income_readable, income))) +
  geom_boxplot(fill = "lightskyblue2") +
  scale_x_continuous(
  name = "Student GPA",
  limits = c(2, 4),
  breaks = c(2, 2.5, 3, 3.5, 4)
) +
  labs(
  title = "Income vs GPA of Students",
  y = "Income Range")
```

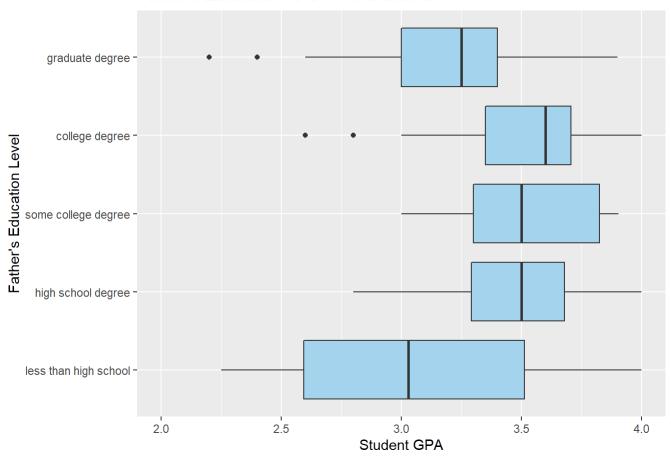
# Income vs GPA of Students



Here we filter out all GPAs that are invalid ( N/A ) and we filter out all father education values that are invalid ( N/A ). We plot the boxplots in order of the father's income labels for ease of readability.

```
food %>%
  filter(
  !is.na(GPA_numeric),
  !is.na(father_education_readable)
  ) %>%
  ggplot(aes(GPA_numeric, fct_reorder(father_education_readable, father_education))) +
  geom_boxplot(fill = "lightskyblue2") +
  scale_x_continuous(
  name = "Student GPA",
  limits = c(2, 4),
  breaks = c(2, 2.5, 3, 3.5, 4)
) +
  labs(
  title = "Father's Education vs GPA of Students",
  y = "Father's Education Level")
```

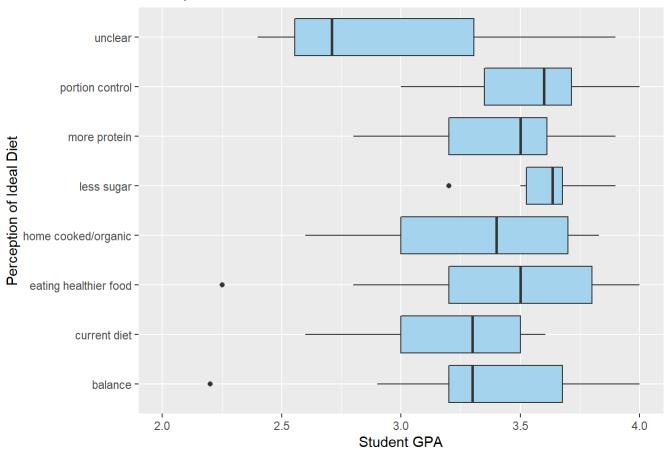
# Father's Education vs GPA of Students



Here we filter out all GPAs that are invalid ( N/A ) and we filter out all ideal diet values that are invalid ( N/A ). Because there is no inherent order to the independent variable, the boxplots are not ordered.

```
food %>%
  filter(
   !is.na(GPA_numeric),
   !is.na(ideal_diet_readable)
   ) %>%
  ggplot(aes(GPA_numeric, ideal_diet_readable)) +
  geom_boxplot(fill = "lightskyblue2") +
  scale_x_continuous(
   name = "Student GPA",
   limits = c(2, 4),
   breaks = c(2, 2.5, 3, 3.5, 4)
) +
  labs(
  title = "Perception of Ideal Diet vs GPA of Students",
   y = "Perception of Ideal Diet")
```

# Perception of Ideal Diet vs GPA of Students



#### **Discussion:**

## Income vs GPA of Students:

From the Income vs GPA of Students plot above, we see that there does not appear to be any significant correlation between the student's income and their GPA. The medians of each range are nearly the same across all categories. There are perhaps a few under-performers in the poorest and richest categories. We could presume this to be either students with difficult life struggles and spoiled students who don't have the need or understand the value of education respectively, but there is not much data to assume this.

### Father's Education vs GPA of Students:

The education of the father of the student does appear to suggest some correlation to the students GPA by the Father's Education vs GPA of Students plot above. The students who were fathered by a high school drop out are statistically more likely to perform worse than their peers. Father's with a graduate degree are also more likely to have a son that does not perform as well in school. We could presume that a high school drop out will not be a responsible father or would not value education.

# Perception of Ideal Diet vs GPA of Students:

As shown from the above plot called Perception of Ideal Diet vs GPA of Students, we can see that most categories do not correlate to the student's GPA. The medians are slightly more varied from the Father's Education vs GPA of Students plot, albeit, with one exception. The unclear category interestingly shows a sharp decrease in student performance. We could presume that students with no strong input are not very organized or responsible or forward thinking. We again do not have the necessary information to conclude any more than this, however.