Beef Analysis - In Class

Load Data and Packages

```
# load packages
pacman::p_load(dplyr, tidyverse, ggplot2, here, readxl, janitor, modelsummary, readr, gridExtra)

# read beef_consumerdata.csv as beef_raw using read.csv() and here() functions
beef_raw <- read.csv(here("data", "beef_consumerdata.csv"))

# create new df called beef that cleans up variable names using clean_names() function
# specify as upper_camel option meaning first letter after any separator (space, period, underscore)
# will be capitalized, e.g., if the variable name initially was "Input.per.capita"
# this function will rename it to InputPerCapita
# we will also rename ConocAmbien to Knoweldge, and BeefConsumptionXTimesPerWeek
# to BeefConsumption

# beef <- beef_raw %>%
# clean_names(., "upper_camel") %>%
# rename(insert_code_here)
```

Data Cleaning and Descriptives

• Tabulate every single variable

```
# insert_code_here
```

- Use mutate() and ifelse() to recode variables where values do not match dictionary
 - recode Knowledge = 2.5 to 3, recode Knowledge = 0 to 1
 - recode Preference1 = 6 to 5
 - recode ChickenConsumption = 8 to 7
- Use mutate() and factor(var, labels = c("label1", "label2")) to convert categorical variables to factor variables for use in analysis later on
 - City and Gen
- Use mutate() and parse_number() to extract numbers from dollar amounts
 - HowMuchMorePerPound (call this var as WtpAmount)
- Use mutate() and ifelse() to recode WtpAmount = 0 if PayMore = 0

```
# beef <- beef %>%
   mutate(Knowledge = insert_code_here,
           Preference1 = insert_code_here,
#
#
           ChickenConsumption = insert code here,
#
           City = insert_code_here,
#
           Gen = insert_code_here,
#
           WtpAmount = insert_code_here,
#
           WtpAmount = insert code here,
#
           Study = factor(Study, labels = c("No education", "Incomplete primary",
                                             "Full primary", "Incomplete secondary",
#
#
                                             "Full secondary", "Technical",
#
                                             "Incomplete university", "University",
#
                                             "Postgraduate")),
           Strata = factor(Strata, labels = c("Strata 1 (lowest)", "Strata 2",
#
#
                                               "Strata 3", "Strata 4", "Strata 5",
#
                                               "Strata 6 (highest)")),
#
           Min12 = factor(Min12, labels = c("No children under 12", "With children under 12")),
           Preference1 = factor(Preference1, labels = c("Beef", "Chicken", "Fish",
#
                                                         "Pork", "Other")),
#
#
           PayMore = factor(PayMore, labels = c("No", "Yes")),
#
           Knowledge = factor(Knowledge, labels = c("No knowledge", "Very little knowledge",
#
                                                     "Some knowledge", "Good knowledge", "Much knowledge
#
           Location = factor(WhereBeef, labels = c("Home", "Out of Home",
#
                                                    "Both Home and Outside")),
           IncomePerCapita = parse_number(InputPerCapita),
#
#
           BeefExpenditurePerCapita = parse_number(BeefExpenditurePerCapita))
```

- Create a new dataframe called beef_expenditure_outlier that contains rows where BeefExpenditurePerCapita $> 99^{th}$ perc
- Create an object called income_exp that contains a scatterplot of BeefExpenditurePerCapita and IncomePerCapita
 - Plot observations where BeefExpenditurePerCapita $> 99^{th}$ percentile as red dots
- Using subset(), remove observations from beef if BeefExpenditurePerCapita $> 99^{th}$ perc
- Create a scatterplot of BeefExpenditurePerCapita and IncomePerCapita without the outliers
- Use grid.arrange() to plot the two scatterplots side by side

```
# create an object called income_exp_removed that contains that scatterplot of the whole dataset

# income_exp_removed <- insert_code_here +

# theme_classic() +

# labs(title = "Removed >99%", y = "Beef expenditure per capita",

# x = "Income per capita")

# use grid.arrange() to plot the two scatterplot objects side by side

# grid.arrange(income_exp, income_exp_removed, ncol = 2)
```

Note: The codes are the same to remove observations if $IncomePerCapita > 99^{th}$ percentile and if $WtpAmount > 99^{th}$ percentile. Use $qeom\ histogram()$ to replicate the qraphs

• Remove outliers based on IncomePerCapita and WtpAmount values and overwrite the beef dataframe

```
# only keep obvs if IncomePerCapita and WtpAmount < 99th perc or if NA
# beef <- beef %>%
# insert_code_here
```

• Use datasummary skim() from the {datasummary} package to generate descriptive stats

```
# descriptive stats for categorical

# beef %>%
# select(City, Gen, Study, Strata, Min12, Preference1, Knowledge, PayMore) %>%
# insert_code_here

# descriptive stats for numerical

# beef %>%
# select(Age, HhSize, BeefConsumption, ChickenConsumption, PorkConsumption, FishConsumption, WtpAmoun
# insert_code_here
```

• Use datasummary_balance() to stratify by a certain variable

```
# stratifying by gender
# datasummary_balance(insert_code_here,
#
                      data = beef %>% select(City, Gen, Study, Strata, Min12,
#
                                              Preference1, Knowledge, PayMore),
#
                      dinm = F,
#
                      title = "Descriptive Statistics of Categorical Variables, Stratified by Gender")
#
# datasummary_balance(insert_code_here,
#
                      data = beef %>%
#
                        select(Gen, Age, HhSize, BeefConsumption, ChickenConsumption,
#
                               PorkConsumption, FishConsumption, WtpAmount,
#
                                IncomePerCapita, BeefExpenditurePerCapita,
#
                               BeefComsumptionPerCapita),
#
                      dinm = F,
#
           title = "Descriptive Statistics of Continuous Variables, Stratified by Gender")
```

Statistical Tests

• Use cor() and cor.test() to calculate correlations

```
# calc correlation between WtpAmount and Age
# insert_code_here
```

• Use t.test() to run t-tests - recode some variables to numeric first

```
# create a new var called GenNumeric that codes 0 if female and 1 if male
# insert_code_here

# t.test if WTP is different for gender (Male = 1)
# insert_code_here
```

- Create a boxplot to look distribution of WtpAmount by Study groups
- Use lm() and anova() to run the ANOVA test, and TukeyHSD(aov()) to determine contrasts
 - Read here for more info

```
# create a boxplot
# insert_code_here +
# theme_classic() +
# theme(axis.text.x = element_text(angle = 45, hjust=1)) +
# labs(x = "Level of Education", y = "WTP Amount")

# run a regression and call it lm_study
# insert_code_here

# print the anova results
# insert_code_here

# print the Tukey Honest Significant Differences test
# insert_code_here
```

Qualtrics API

Instead of downloading your survey every time from the Qualtrics website, you can connect directly to the server with an API using the {qualtRics} package.

You will need an API key.

• In your UBC Qualtrics Account, go to **Account Settings** -> **Qualtrics IDs** tab -> on the left under the API table click **Generate Token**. Copy the **Token** and paste it in the code below under api_key.

You will need the **survey id** of your survey.

• In your UBC Qualtrics Account, go to **Account Settings** -> **Qualtrics IDs** tab -> on the right under the *Surveys* table copy the code beside the name of your survey and paste it in the code below under **SurveyID**. On my end, my surveys all start with SV_, but I'm not sure if that's true for all users.

```
pacman::p_load(qualtRics)

# Establish API
# qualtrics_api_credentials(api_key = "insert_code_here",
# base_url = "ubc.ca1.qualtrics.com",
# install = TRUE)

# Load your survey data to a dataframe called rawsurveydata
# rawsurveydata <- fetch_survey(surveyID = "insert_code_here")</pre>
```