

assignment_02

August 4, 2025

1 Assignment 2

1. Multivariate Data Analysis & Vizualisation (11)

For the `vehicles` dataset from the `fueleconomy` package in R,

- a. develop the following plots (10)
 - histogram for highway & city MpG
 - box plot for highway & city MpG
 - bar plot for mean highway & city MpG vs. vehicle year
 - line plot for mean highway & city MpG vs. engine cylinders
 - scatter plot for highway & city MpG vs. engine displacement
- b. evaluate the correlation between (1)
 - engine displacement and highway & city MpG
 - engine cylinders and highway & city MpG
 - vehicle year and highway & city MpG

```
[ ]: # Vehicles Dataset - Fuel Economy Package
## Load Packages
library(dplyr)
library(ggplot2)
library(patchwork)
library(fueleconomy)

# Load the dataset (filtered for non-zero & non-NA values; augmented with new
  ↪variable 'I' indicating if vehicle is manufactured after 2009)
data <- fueleconomy::vehicles %>%
  filter(year != 0, cyl != 0, displ != 0, is.na(year) == FALSE, is.na(cyl) ==
  ↪FALSE, is.na(displ) == FALSE) %>%
  mutate(I = ifelse(year > 2009, 1, 0))
```

2. Linear Regression (18)

For the `vehicles` dataset from the `fueleconomy` package in R,

- a. develop the following models (5)
 - $\text{cty} = \beta_0 + \beta_1 \text{year} + \beta_2 \text{cyl} + \beta_3 X_3 \text{ displ}$
 - $\text{hwy} = \beta_0 + \beta_1 \text{year} + \beta_2 \text{cyl} + \beta_3 X_3 \text{ displ}$
- b. for the two models, compute (6)
 - Sum of Squares Total (SST)
 - Sum of Squares Regression (SSR)
 - Sum of Squared Errors (SSE)
 - Residual Standard Error (RSE)
 - R-squared (R^2)
 - Adjusted R-squared (\bar{R}^2)
- c. for the two models, (7)
 - develop residuals plot
 - comment upon the validity of the assumptions of linear regression

(Hint: To comment upon multicollinearity, develop pairwise correlation for the exogenous variables)

3. Symbolic Regression (5)

For the `vehicles` dataset from the `fueleconomy` package in R, compare the following models explored via symbolic regression with the linear regression model developed in the previous question (5)

- $\text{cty} = \beta_0 + \beta_1 \text{I}(\text{year} \geq 2010) + \beta_2 \log(\text{displ})$
- $\text{hwy} = \beta_0 + \beta_1 \text{I}(\text{year} \geq 2010) + \beta_2 \log(\text{displ})$

(Hint: Compare the two set of models using model statistics and residual plots)

4. Logistic Regression (11)

For the `TravelMode` dataset from the `AER.` package in R, explore how alternate-specific variables (`travel`: in-vehicle travel time, `wait`: terminal waiting time, `vcost`: vehicle operational cost, and `gcost`: generalized travel cost) as well as individual-specific variables (`income`: household income, and `size`: traveling party size) impact choice of travel mode (`air`, `train`, `bus`, `car`). To this end,

(Note: The data is available in **long format** with one row per *individual-mode* combination.)

- a. develop the model (4)
- b. compute the following statistics (7)
 - log-likelihood for the
 - equally likely model

- market share model
- estimated model
- estimated model R-squared with respect to the
 - equally-likely model
 - market share model
- estimated model adjusted R-squared with respect to the
 - equally-likely model
 - market share model

```
[ ]: # TravelMode Dataset - AER Package
## Load Packages
library(AER)
library(dplyr)
library(mlogit)

## Load the dataset (choices mutated from yes/no to TRUE/FALSE)
data("TravelMode", package = "AER")
long_data <- TravelMode %>%
  mutate(choice = choice == "yes")

model_data <- mlogit.data(
  long_data,
  choice = "choice",
  shape = "long",
  chid.var = "individual",
  alt.var = "mode"
)
```