STA302 Proposal

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load the library

```
library(tidyverse)
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr 1.1.2 v readr 2.1.4
## v forcats 1.0.0 v stringr 1.5.0
## v ggplot2 3.4.2 v tibble 3.2.1
                                  1.3.0
## v lubridate 1.9.2
                       v tidyr
             1.0.1
## v purrr
## -- Conflicts -----
                                      ------tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
library(knitr)
library(dplyr)
# set wd
# setwd("~/Desktop/Github Projects/STA302")
```

extract the subset of data we need

```
coffee_df <- read.csv(file = "merged_data_cleaned.csv") %>%
  rename("Ratings" = Total.Cup.Points) %>%
  mutate(.before = 1, InSpecies = ifelse(Species == "Arabica", 1, 0)) %>%
  select(Species, InSpecies, Aroma, Flavor, Aftertaste, Acidity, Sweetness, Ratings)
```

numerical summary

```
summary(coffee_df)
```

```
## Length:1339
                      Min. :0.0000
                                     Min.
                                            :0.000 Min.
                                                             :0.00
## Class:character 1st Qu.:1.0000
                                      1st Qu.:7.420 1st Qu.:7.33
## Mode :character Median :1.0000
                                     Median :7.580
                                                     Median:7.58
##
                      Mean
                            :0.9791
                                     Mean
                                            :7.567
                                                      Mean
                                                             :7.52
##
                      3rd Qu.:1.0000 3rd Qu.:7.750 3rd Qu.:7.75
##
                      Max.
                            :1.0000 Max. :8.750 Max.
                                                             :8.83
##
     Aftertaste
                      Acidity
                                    Sweetness
                                                      Ratings
## Min.
          :0.000 Min.
                          :0.000
                                  Min. : 0.000 Min.
                                                          : 0.00
##
  1st Qu.:7.250 1st Qu.:7.330
                                  1st Qu.:10.000 1st Qu.:81.08
## Median :7.420 Median :7.580
                                  Median :10.000
                                                  Median :82.50
## Mean :7.401
                   Mean :7.536
                                  Mean : 9.857
                                                   Mean
                                                          :82.09
## 3rd Qu.:7.580
                   3rd Qu.:7.750
                                  3rd Qu.:10.000
                                                   3rd Qu.:83.67
## Max. :8.670 Max. :8.750
                                  Max. :10.000
                                                   Max. :90.58
summary_table <- coffee_df %>%
  summarise(
    Aroma_mean = mean(Aroma),
   Flavor_mean = mean(Flavor),
   Aftertaste_mean = mean(Aftertaste),
   Sweetness_mean = mean(Sweetness),
   Acidity_mean = mean(Acidity),
   Aroma_sd = sd(Aroma),
   Flavor_sd = sd(Flavor),
   Aftertaste_sd = sd(Aftertaste),
   Sweetness_sd = sd(Sweetness),
   Acidity_sd = sd(Acidity),
   Aroma_max = max(Aroma),
   Flavor_max = max(Flavor),
   Aftertaste_max = max(Aftertaste),
   Sweetness_max = max(Sweetness),
   Acidity_max = max(Acidity),
   Aroma_min = min(Aroma),
   Flavor_min = min(Flavor),
   Aftertaste_min = min(Aftertaste),
   Sweetness_min = min(Sweetness),
    Acidity_min = min(Acidity),
)
summary_frame = data.frame(
  Variables = c("Flavor", "Aroma", "Sweetness", "Aftertaste", "Acidity"),
 Min = c(summary_table$Flavor_min, summary_table$Aroma_min, summary_table$Sweetness_min, summary_table
 Max = c(summary_table$Flavor_max, summary_table$Aroma_max, summary_table$Sweetness_max, summary_table
 Mean = c(summary_table$Flavor_mean, summary_table$Aroma_mean, summary_table$Sweetness_mean, summary_t
  SD = c(summary_table$Flavor_sd, summary_table$Aroma_sd, summary_table$Sweetness_sd, summary_table$Aft
kable(summary_frame, format = "markdown", caption = "Coffee Ratings Dataset Numerical Summary",
      col.names = c("Variable Name", "Minimum", "Maximum", "Mean", "Standard Deviation"),
     align = "c", longtable = TRUE, digits = 3)
```

Aroma

Flavor

Species

InSpecies

##

Table 1: Coffee Ratings Dataset Numerical Summary

Variable Name	Minimum	Maximum	Mean	Standard Deviation
Flavor	0	8.83	7.520	0.398
Aroma	0	8.75	7.567	0.378
Sweetness	0	10.00	9.857	0.616
Aftertaste	0	8.67	7.401	0.404
Acidity	0	8.75	7.536	0.380

```
coffee_df %>% group_by(Species) %>%
  summarise(num_species = n()) %>%
  rename(`Number of Observations` = num_species) %>%
  kable(align = "c", longtable = TRUE)
```

Species	Number of Observations
Arabica	1311
Robusta	28

The coffee ratings dataset contains 1339 observations, with 1311 of these being of the "Arabica" species, while the other 28 are of the "Robusta" species. The minimum for all variables is 0, while the maximum rating is given in the Sweetness category. Sweetness also has the highest mean and standard deviation.

Fit the model

```
model <- lm(Ratings ~ InSpecies + Aroma + Flavor + Aftertaste + Acidity + Sweetness, data = coffee_df)
model
##
## Call:
## lm(formula = Ratings ~ InSpecies + Aroma + Flavor + Aftertaste +
       Acidity + Sweetness, data = coffee_df)
##
##
## Coefficients:
## (Intercept)
                  InSpecies
                                    Aroma
                                                Flavor
                                                          Aftertaste
                                                                           Acidity
##
         7.166
                     -2.530
                                    1.364
                                                  2.338
                                                               2.525
                                                                             1.272
##
     Sweetness
##
         2.153
```

assumption checking

```
# fitted values
y_hat <- fitted(model)
# residual
e_hat <- resid(model)</pre>
```

```
# attached these columns to coffee_df
coffee_df <- coffee_df %>%
  mutate("y_hat" = fitted(model)) %>%
  mutate("e_hat" = resid(model))
# write.csv(coffee_df, "CoffeeRatings.csv", row.names = FALSE)
```

residual-fitted value plot

Residual vs Fitted

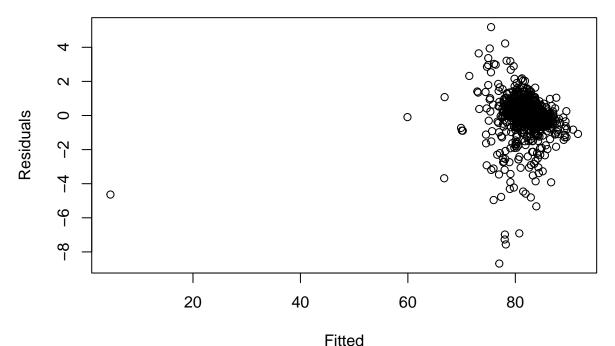


Fig. 1: The residuals versus fitted values plot from the model.

```
# plot1 <-
# ggplot(data = coffee_df,
# mapping = aes(x = y_hat, y = e_hat))+
# geom_point(alpha = 0.5, size = 2)+
# theme_bw()+
# theme(plot.title = element_text(hjust = 0.5))+</pre>
```

```
# labs(x = "Fitted", y = "Residuals", title = "Residual vs. Fitted")
# ggsave(plot1, file = "ResidFitted.png", dpi = 600, width = 5, height = 5)
```

all residual vs. predictors

```
par(mfrow = c(2, 3), oma = c(4, 2, 0, 0) + 0.1)
plot(x = coffee_df$InSpecies, y = e_hat,
                      main = "Residual vs. Species", xlab = "Species",
                      ylab = "Residual")
plot(x = coffee_df$Aroma, y = e_hat,
                    main = "Residual vs. Aroma", xlab = "Aroma",
                    ylab = "Residual")
plot(x = coffee_df$Flavor, y = e_hat,
                     main = "Residual vs. Flavor", xlab = "Flavor",
                     ylab = "Residual")
plot(x = coffee_df$Aftertaste, y = e_hat,
                    main = "Residual vs. Aftertaste", xlab = "Aftertaste",
                    ylab = "Residual")
plot(x = coffee_df$Acidity, y = e_hat,
                      main = "Residual vs. Acidity", xlab = "Acidity",
                      ylab = "Residual")
plot(x = coffee_df$Sweetness, y = e_hat,
                    main = "Residual vs. Sweetness", xlab = "Sweetness",
                    ylab = "Residual")
mtext(~italic("Fig.4: All residuals versus predictors plots from the model."), side = 1, line = 3, oute
```

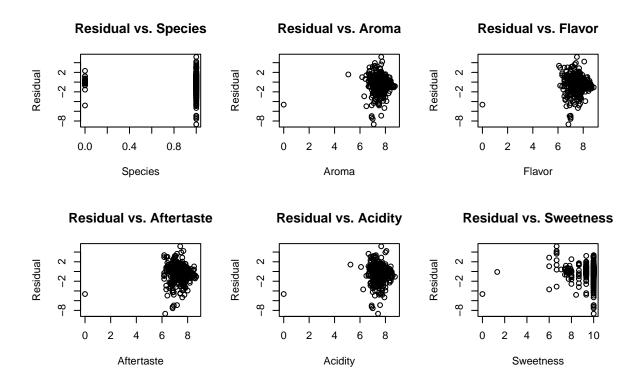


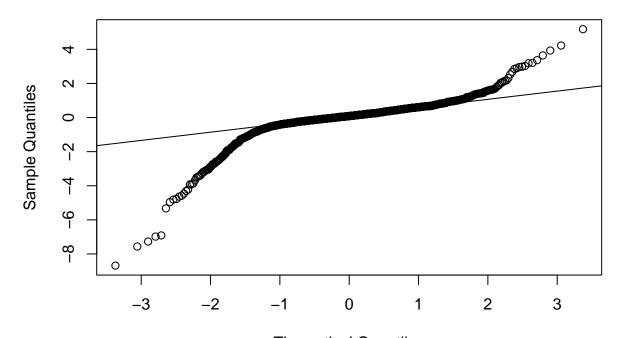
Fig.4: All residuals versus predictors plots from the model.

```
# title(sub = "Figure4: All residuals versus predictors from the model.", adj = 0.5)
```

checking normality of errors

```
qqnorm(e_hat)
qqline(e_hat)
title(sub = ~italic("Fig.3: The normal Q-Q plot of residues from the model."))
```

Normal Q-Q Plot



Theoretical Quantiles Fig.3: The normal Q–Q plot of residues from the model.

Response vs. Fitted

Response vs. Fitted

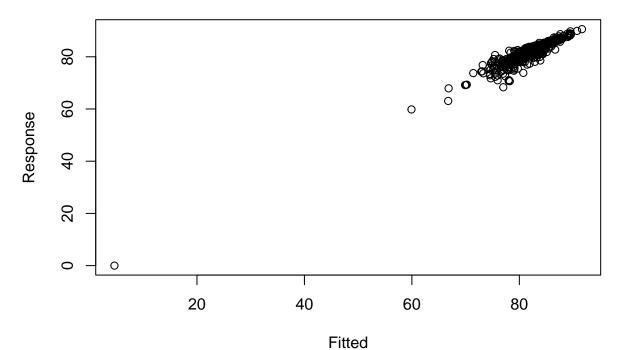


Fig.2: The reponse (coffee ratings) versus fitted values plot from the model.

pair plots of every predictors

```
pairs(coffee_df[, c(2:7)])
title(sub = ~italic("Fig.5: The pairwise scatter plots of all predictors."))
```

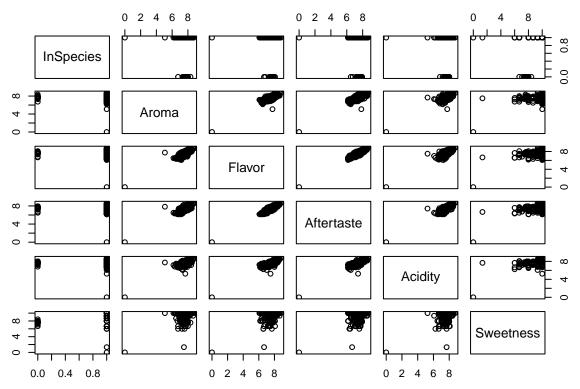


Fig.5: The pairwise scatter plots of all predictors.