Lab 3

Math 241, Week 3

```
libs <- c('tidyverse', 'knitr', 'viridis', 'mosaic', 'mosaicData', 'babynames', 'Lahman', 'nycflights13', 'rn
for(l in libs){
   if(!require(1, character.only = TRUE, quietly = TRUE)){
     message( sprintf('Did not have the required package << %s >> installed. Downloading now ... ',l))
     install.packages(l)
   }
   library(l, character.only = TRUE, quietly = TRUE)
}
```

Due: Friday, February 16th at 8:30am

Goals of this lab

- 1. Practice creating functions.
- 2. Practice refactoring your code to make it better! Therefore for each problem, make sure to test your functions.

Problem 1: Subset that R Object

Here are the R objects we will use in this problem (dats, pdxTreesSmall and ht).

a. What are the classes of dats, pdxTreesSmall and ht?

```
class(dats)
## [1] "list"
```

```
class(pdxTreesSmall)
## [1] "tbl_df"
                                    "data.frame"
                      "tbl"
class(ht)
## [1] "numeric"
  b. Find the 10th, 11th, and 12th values of ht.
ht[10:12]
## [1] 112 112 48
  c. Provide the Species column of pdxTrees as a data frame with one column.
pdxTrees_species_df <- select(pdxTrees, Species)</pre>
pdxTrees_species_df
## # A tibble: 25,534 x 1
##
      Species
##
      <chr>
##
    1 PSME
    2 PSME
##
    3 CRLA
##
   4 QURU
##
    5 PSME
##
    6 PSME
    7 PSME
##
##
    8 PSME
## 9 PSME
## 10 PSME
## # i 25,524 more rows
  d. Provide the Species column of pdxTrees as a character vector.
pdxTrees_species_chr <- head(pdxTrees$Species)</pre>
as.character(pdxTrees_species_chr)
## [1] "PSME" "PSME" "CRLA" "QURU" "PSME" "PSME"
  e. Provide code that gives us the second entry in sets from dats.
dats$sets[2]
```

[1] "Births2015"

f. Subset pdxTreesSmall to only Douglas-fir and then provide the DBH and Condition of the 4th Douglas-fir in the dataset. (Feel free to mix in some tidyverse code if you would like to.)

```
library(dplyr)

pdxTreesSmall %>%
  filter(Species == "Douglas-fir") %>%
  slice(4) %>%
  select(DBH, Condition)
```

```
## # A tibble: 0 x 2
## # i 2 variables: DBH <dbl>, Condition <chr>
```

Problem 2: Function Creation

Figure out what the following code does and then turn it into a function. For your new function, do the following:

- Test it.
- Provide default values (when appropriate).
- Use clear names for the function and arguments.
- Make sure to appropriately handle missingness.
- Generalize it by allowing the user to specify a confidence level.
- Check the inputs and stop the function if the user provides inappropriate values.

```
library(pdxTrees)
thing1 <- length(pdxTrees$DBH)
thing2 <- mean(pdxTrees$DBH)
thing3 <- sd(pdxTrees$DBH)/sqrt(thing1)
thing4 <- qt(p = .975, df = thing1 - 1)
thing5 <- thing2 - thing4*thing3
thing6 <- thing2 + thing4*thing3</pre>
```

```
calculate_CI <- function(data, confidence_level = 0.95) {
   if (missing(data) || is.null(data)) {
      stop("Please provide a valid data set.")
   }

if (confidence_level <= 0 || confidence_level >= 1) {
      stop("Confidence level must be between 0 and 1.")
   }

n <- length(data)
mean_val <- mean(data)
std_err <- sd(data) / sqrt(n)
t_value <- qt(p = (1 + confidence_level) / 2, df = n - 1)

lower_bound <- mean_val - t_value * std_err
upper_bound <- mean_val + t_value * std_err</pre>
```

```
result <- list(</pre>
    mean = mean_val,
    lower_bound = lower_bound,
    upper_bound = upper_bound,
    confidence_level = confidence_level
  return(result)
}
result <- calculate_CI(pdxTrees$DBH, confidence_level = 0.95)</pre>
print(result)
## $mean
## [1] 20.61408
##
## $lower_bound
## [1] 20.44981
##
## $upper_bound
## [1] 20.77835
##
```

Problem 3: Wrapper Function for your ggplot

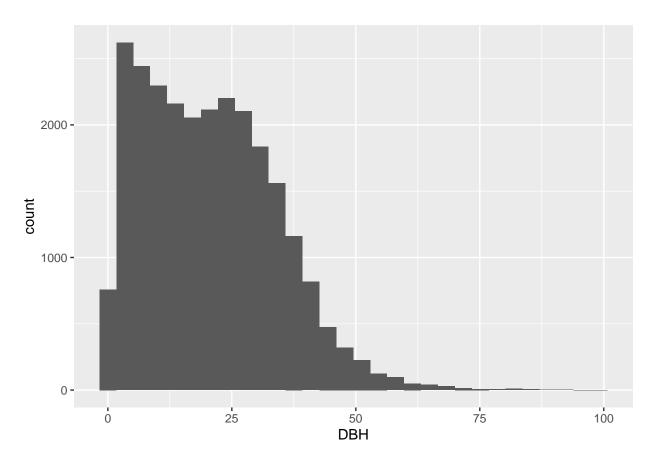
\$confidence_level

[1] 0.95

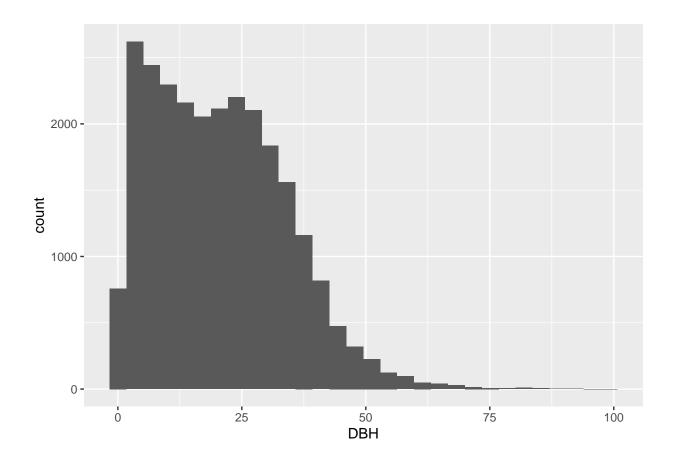
While we (i.e. Math 241 students) all love the grammar of graphics, not everyone else does. So for this problem, we are going to practice creating wrapper functions for ggplot2.

Here's an example of a wrapper for a histogram. Notice that I can't just list the variable name as an argument. The issue has to do with how many of the tidyverse functions evaluate the arguments. Therefore we have to quote (enquo()) and then unquote (!!) the arguments. (If you want to learn more, go here.)

```
# Minimal viable product working code
ggplot(data = pdxTrees, mapping = aes(x = DBH)) +
geom_histogram()
```



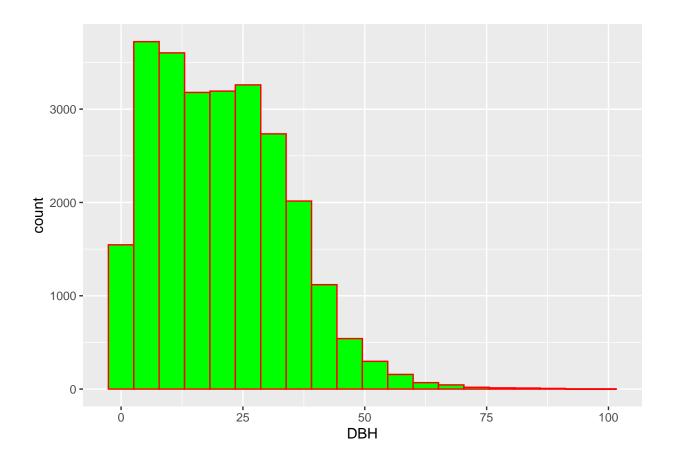
```
# Shorthand histogram function
histo <- function(data, x){
    x <- enquo(x)
    ggplot(data = data, mapping = aes(x = !!x)) +
        geom_histogram()
}
# Test it
histo(pdxTrees, DBH)</pre>
```



- a. Edit histo() so that the user can set
- The number of bins
- $\bullet\,$ The fill color for the bars
- $\bullet~$ The color outlining the bars

```
histo <- function(data, x, bins = 30, fill_color = "blue", outline_color = "black") {
    x <- enquo(x)
    ggplot(data = data, mapping = aes(x = !!x)) +
        geom_histogram(bins = bins, fill = fill_color, color = outline_color)
}

# Test it with custom parameters
histo(pdxTrees, DBH, bins = 20, fill_color = "green", outline_color = "red")</pre>
```

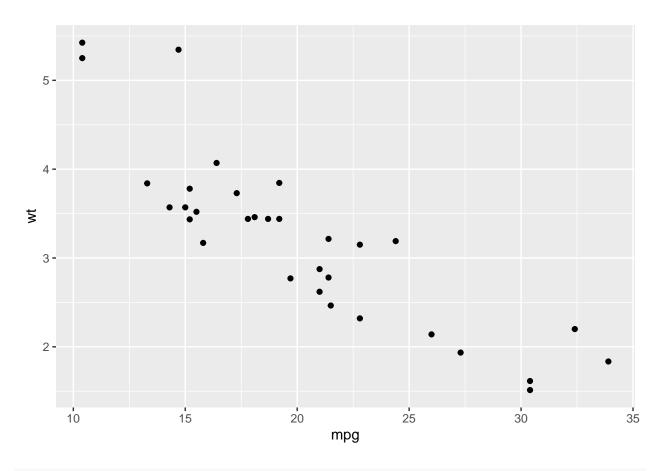


b. Write code to create a basic scatterplot with ggplot2. Then write and test a function to create a basic scatterplot.

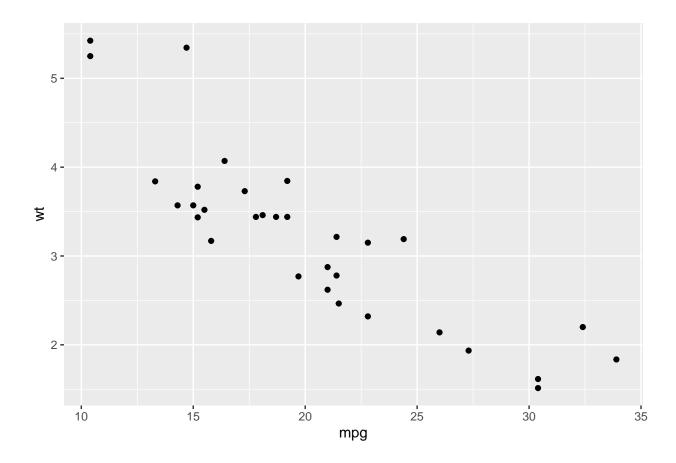
```
library(ggplot2)

basic_scatterplot <- ggplot(data = mtcars, aes(x = mpg, y = wt)) +
    geom_point()

print(basic_scatterplot)</pre>
```

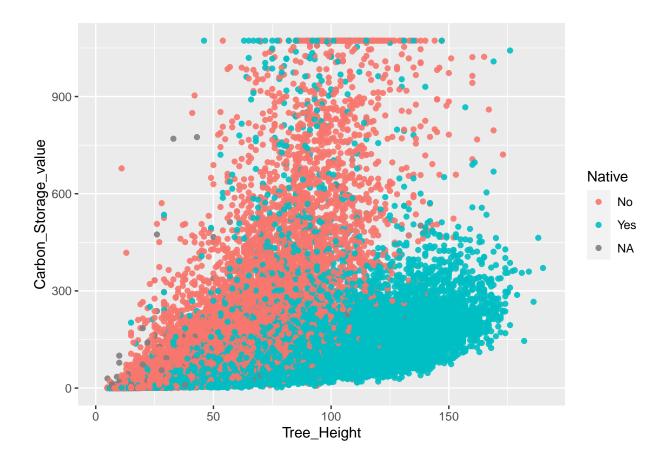


```
scatterplot_basic <- function(data, x, y) {
  ggplot(data = data, aes(x = {{ x }}, y = {{ y }})) +
     geom_point()
}
scatterplot_basic(mtcars, mpg, wt)</pre>
```



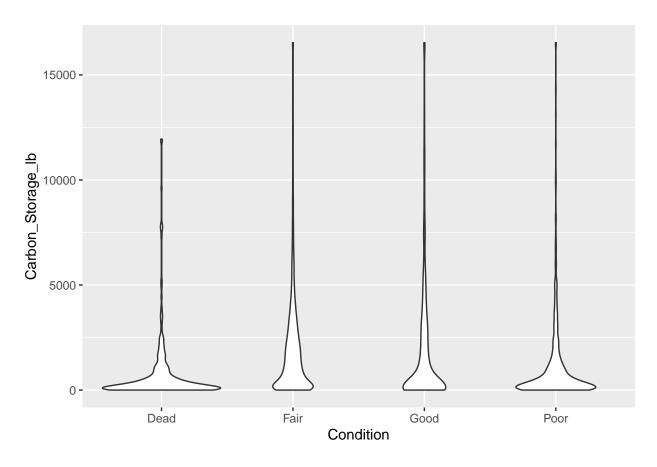
- c. Modify your scatter plot function to allow the user to \dots
- $\bullet\,$ Color the points by another variable.
- $\bullet~$ Set the transparency.

```
scatter_function <- function(data, x, y, alpha = 0.5, color = Condition){
    x <- enquo(x)
    y <- enquo(y)
    color <- enquo(color)
    ggplot(data = data, mapping = aes(x = !!x, y = !!y, color = !!color)) +
        geom_point(alpha = alpha)
}
# Testing scatter function
scatter_function(pdxTrees, Tree_Height, Carbon_Storage_value, alpha = 0.9, color = Native)</pre>
```

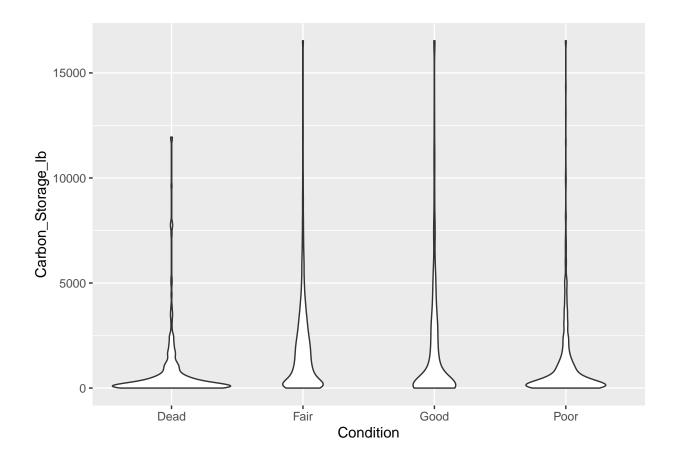


d. Write and test a function for your favorite ggplot2 graph.

```
#Basic violin plot code
ggplot(data = pdxTrees, aes(x = Condition, y = Carbon_Storage_lb)) +
    geom_violin()
```



```
#Function for basic violin plot
violin_function <- function(data, x, y){
    x <- enquo(x)
    y <- enquo(y)
    ggplot(data = data, mapping = aes(x = !!x, y = !!y)) +
        geom_violin()
}
#Testing violin function
violin_function(pdxTrees, Condition, Carbon_Storage_lb)</pre>
```



Problem 4: Functioning dplyr

a. Take the following code and turn it into an R function to create a **conditional proportions** table. Similar to ggplot2, you will need to quote and unquote the variable names. Make sure to test your function!

```
pdxTrees %>%
  count(Native, Condition) %>%
  group_by(Native) %>%
  mutate(prop = n/sum(n)) %>%
  ungroup()
```

```
## # A tibble: 10 x 4
##
      Native Condition
                            n
                                  prop
##
      <chr>
             <chr>
                        <int>
                                 <dbl>
##
    1 No
             Fair
                        12284 0.865
##
    2 No
             Good
                         1043 0.0734
##
    3 No
             Poor
                          875 0.0616
##
    4 Yes
             Fair
                         9877 0.904
    5 Yes
             Good
                          600 0.0549
##
    6 Yes
             Poor
                          454 0.0415
##
    7 <NA>
             Dead
                          264 0.658
##
    8 <NA>
             Fair
                          118 0.294
    9 <NA>
             Good
                            3 0.00748
                           16 0.0399
## 10 <NA>
             Poor
```

```
conditional_proportions_table <- function(data, var1, var2) {
   data %>%
      count(!!enquo(var1), !!enquo(var2)) %>%
      group_by(!!enquo(var1)) %>%
      mutate(prop = n/sum(n)) %>%
      ungroup()
}

result <- conditional_proportions_table(pdxTrees, Native, Condition)
print(result)</pre>
```

```
## # A tibble: 10 x 4
##
     Native Condition
                        n
                              prop
     <chr> <chr>
##
                     <int>
                             <dbl>
## 1 No
            Fair
                     12284 0.865
## 2 No
                     1043 0.0734
            Good
## 3 No
            Poor
                       875 0.0616
## 4 Yes
            Fair
                      9877 0.904
## 5 Yes
            Good
                       600 0.0549
## 6 Yes
            Poor
                       454 0.0415
## 7 <NA>
                       264 0.658
            Dead
## 8 <NA>
            Fair
                       118 0.294
## 9 <NA>
            Good
                         3 0.00748
## 10 <NA>
            Poor
                        16 0.0399
```

b. Write a function to compute the mean, median, sd, min, max, sample size, and number of missing values of a quantitative variable by the categories of another variable. Make sure the output is a data frame (or tibble). Don't forget to test your function.

```
library(dplyr)
quant_summary_by_category <- function(data, category_var, quant_var) {</pre>
  data %>%
    group_by(!!enquo(category_var)) %>%
    summarize(
      mean = mean(!!enquo(quant_var), na.rm = TRUE),
      median = median(!!enquo(quant_var), na.rm = TRUE),
      sd = sd(!!enquo(quant_var), na.rm = TRUE),
      min = min(!!enquo(quant_var), na.rm = TRUE),
      max = max(!!enquo(quant_var), na.rm = TRUE),
      n = n(),
      missing_values = sum(is.na(!!enquo(quant_var)))
    ) %>%
    ungroup()
}
example_data <- data.frame(</pre>
  Category = rep(c("A", "B", "C"), each = 5),
  Value = c(1, 2, 3, 4, NA, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14)
)
```

```
result <- quant_summary_by_category(example_data, Category, Value)
print(result)</pre>
```

```
## # A tibble: 3 x 8
     Category mean median
                                                   n missing_values
                               sd
                                    min
                                           max
                                                               <int>
##
                      <dbl> <dbl> <dbl> <int>
     <chr>>
              <dbl>
## 1 A
                        2.5
                            1.29
                                             4
                                                   5
                                                                   1
                 2.5
                                       1
                7
                                                                   0
## 2 B
                        7
                             1.58
                                       5
                                             9
                                                   5
## 3 C
               12
                       12
                             1.58
                                      10
                                            14
                                                   5
                                                                   0
```

Problem 5: another babynames exercise

Write a function called grab_name that, when given a **name** and a year as an argument, returns the rows from the babynames data frame in the babynames package that match that name for that year (and returns an error if that name and year combination does not match any rows). Run the function once with the arguments **Ezekiel and 1883** and once with **Ezekiel and 1983**.

```
grab_name <- function(myname, myyear) {</pre>
  result <- babynames %>%
    filter(name == myname, year == myyear)
  if (nrow(result) == 0) {
    stop("No matching rows found for the given name and year combination.")
  return(result)
}
# Test with Ezekiel and 1883
result_1883 <- grab_name("Ezekiel", 1883)</pre>
print(result_1883)
## # A tibble: 1 x 5
##
      year sex
                 name
                                     prop
##
     <dbl> <chr> <chr>
                                    <dbl>
                          <int>
## 1 1883 M
                  Ezekiel
                             14 0.000124
# Test with Ezekiel and 1983
result_1983 <- grab_name("Ezekiel", 1983)</pre>
print(result_1983)
## # A tibble: 1 x 5
##
      year sex
                 name
                              n
                                      prop
     <dbl> <chr> <chr>
                          <int>
                                     <dbl>
## 1 1983 M
                 Ezekiel
                            149 0.0000800
```