

#1. Use the dataset mpg

```
datmpg <- read.csv("mpg.csv")
```

#1b. Which variables from mpg dataset are categorical?

```
categorical_vars <- c("manufacturer", "model", "trans", "drv", "fl", "class")
categorical_vars
```

```
## [1] "manufacturer" "model"          "trans"          "drv"            "fl"
## [6] "class"
```

#1c. Which are continuous variables?

```
continuous_vars <- c("displ", "year", "cyl", "cty", "hwy")
continuous_vars
```

```
## [1] "displ" "year"  "cyl"   "cty"   "hwy"
```

#2. Which manufacturer has the most models in this data set? Which model has the most variations? Show your codes and result.

#2a. Group the manufacturers and find the unique models. Show your codes and result.

```
library(dplyr)
```

```
##
```

```
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
## filter, lag
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
## intersect, setdiff, setequal, union
```

```
manufacturer_models <- datmpg %>%
  group_by(manufacturer) %>%
  summarise(num_models = n_distinct(model)) %>%
  arrange(desc(num_models))
```

```
manufacturer_models[1, ]
```

```
## # A tibble: 1 x 2
```

```
##   manufacturer num_models
```

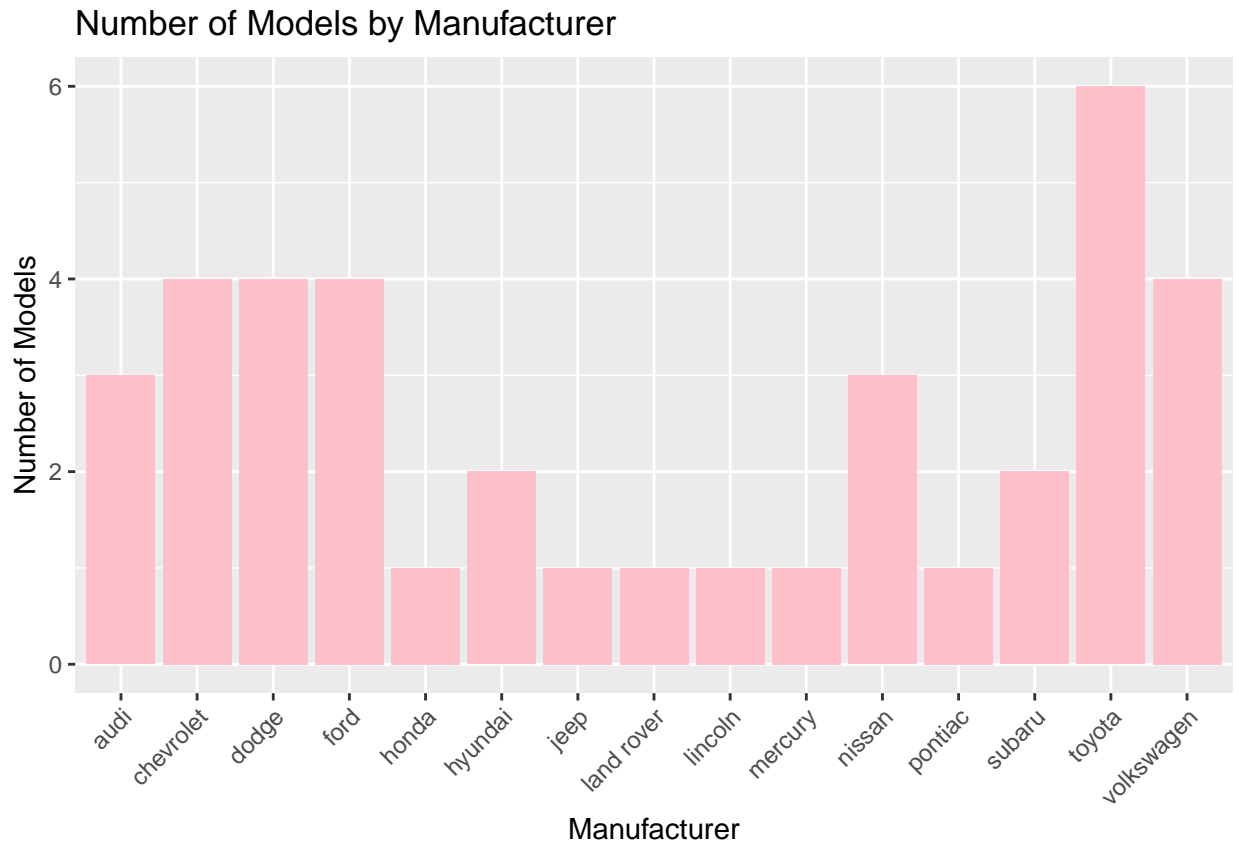
```
##   <chr>          <int>
```

```
## 1 toyota          6
```

#2b. Graph the result by using plot() and ggplot(). Write the codes and its result

```
library(ggplot2)
```

```
ggplot(manufacturer_models, aes(x = manufacturer, y = num_models)) +
  geom_bar(stat = "identity", fill = "pink") +
  labs(title = "Number of Models by Manufacturer", x = "Manufacturer", y = "Number of Models") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



#2.

#2a. What does `ggplot(mpg, aes(model, manufacturer)) + geom_point()` show?

#This `ggplot` command creates a scatter plot where each point represents a car model, positioned along the x-axis by manufacturer and the y-axis by model.

#2b. Is it useful? If not, how could you modify the data to make it more informative?

#The plot may be useful for visualizing the distribution of models across manufacturers, but it could be improved by adding more context, such as color-coding by year or engine type.

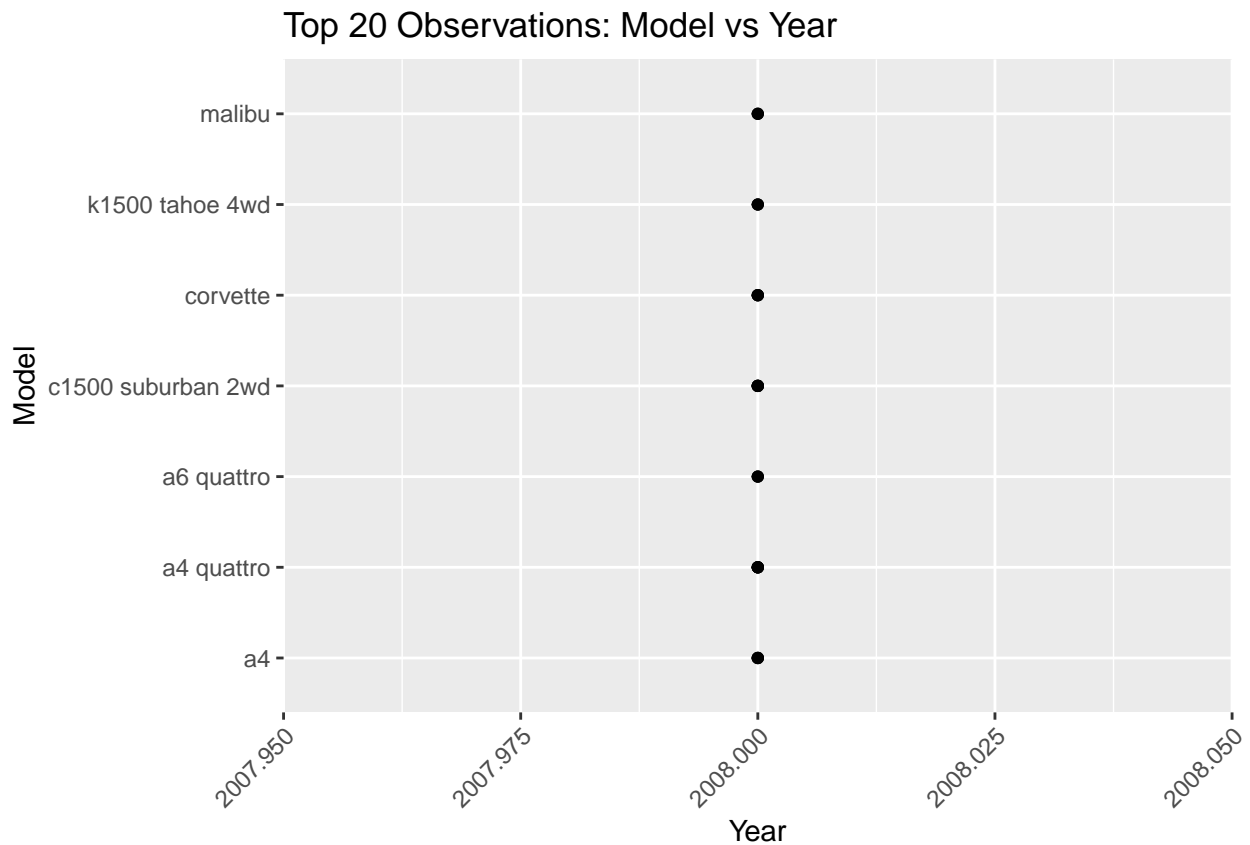
#3.

```
top_20_data <- head(mpg[order(mpg$year, decreasing = TRUE), ], 20)
```

```
ggplot(top_20_data, aes(x = year, y = model)) +
```

```
  geom_point() +
```

```
  labs(title = "Top 20 Observations: Model vs Year", x = "Year", y = "Model") + theme(axis.text.x = element_text(angle = 45))
```



#4. Using the pipe (%>%), group the model and get the number of cars per model. Show codes and its results.

```
library(dplyr)

cars_per_model <- mpg %>%
  group_by(model) %>%
  summarise(num_cars = n())

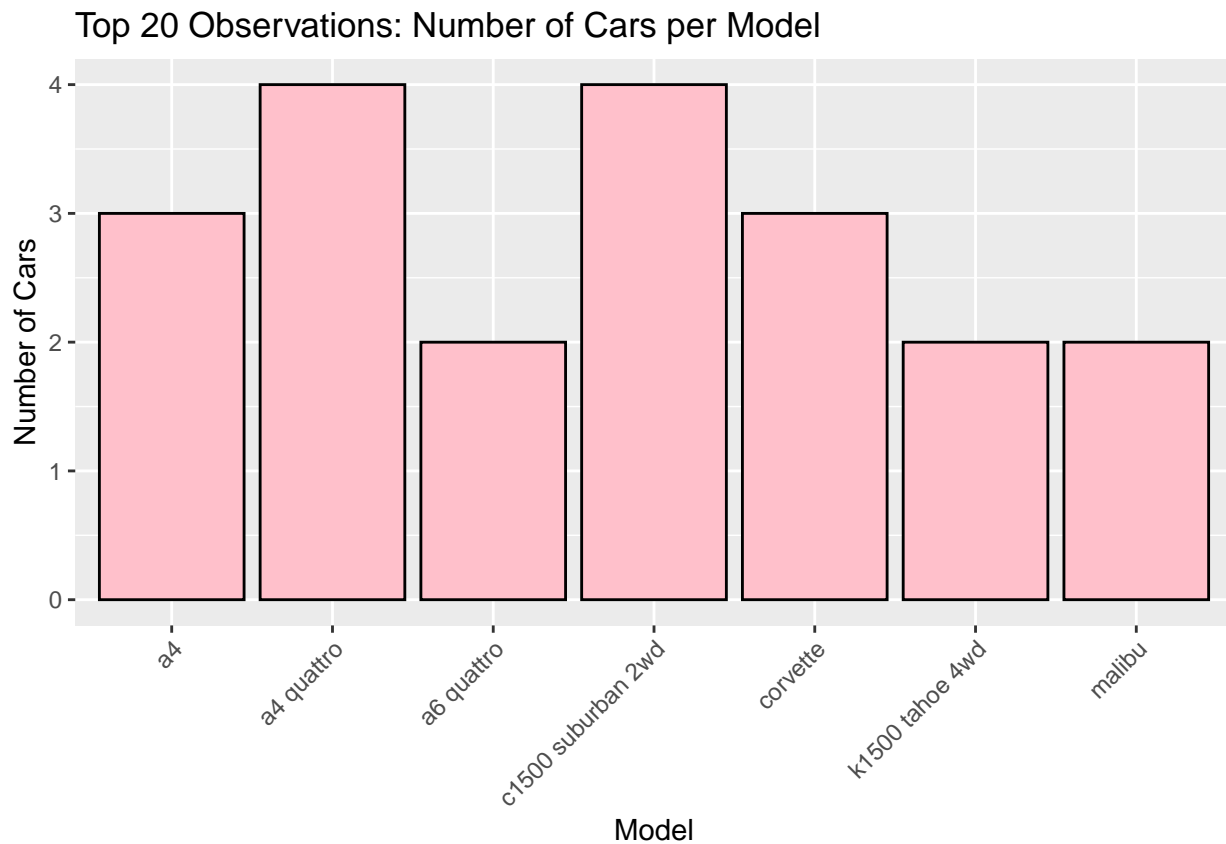
print(cars_per_model)
```

```
## # A tibble: 38 x 2
##   model          num_cars
##   <chr>          <int>
## 1 4runner 4wd             6
## 2 a4                     7
## 3 a4 quattro             8
## 4 a6 quattro             3
## 5 altima                 6
## 6 c1500 suburban 2wd     5
## 7 camry                  7
## 8 camry solara           7
## 9 caravan 2wd           11
## 10 civic                  9
## # i 28 more rows
```

#4a. Plot using geom_bar() using the top 20 observations only. The graphs should have a title, labels and axis titles.

```
top_20_data <- head(mpg[order(mpg$year, decreasing = TRUE), ], 20)
```

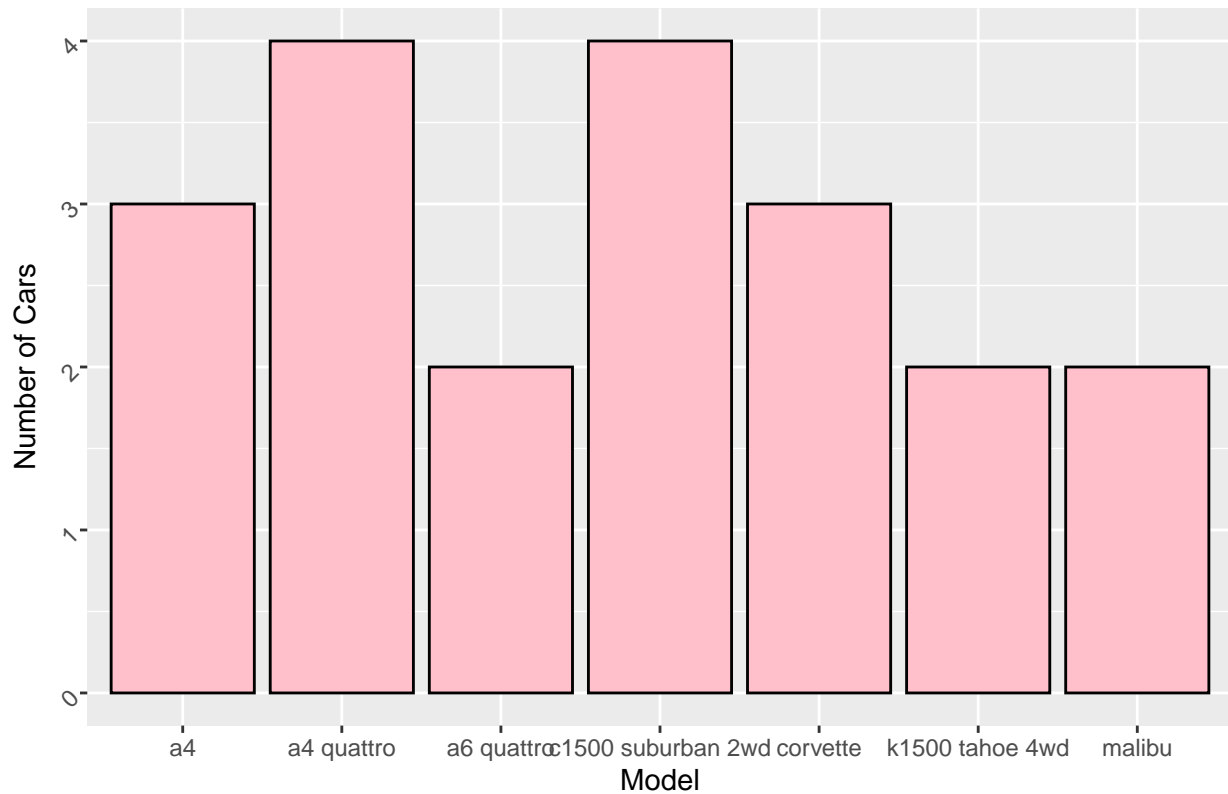
```
ggplot(top_20_data, aes(x = model)) +
  geom_bar(fill = "pink", color = "black") +
  labs(title = "Top 20 Observations: Number of Cars per Model",
       x = "Model", y = "Number of Cars") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



#4b. Plot using the `geom_bar()` + `coord_flip()` just like what is shown below. Show codes and its result

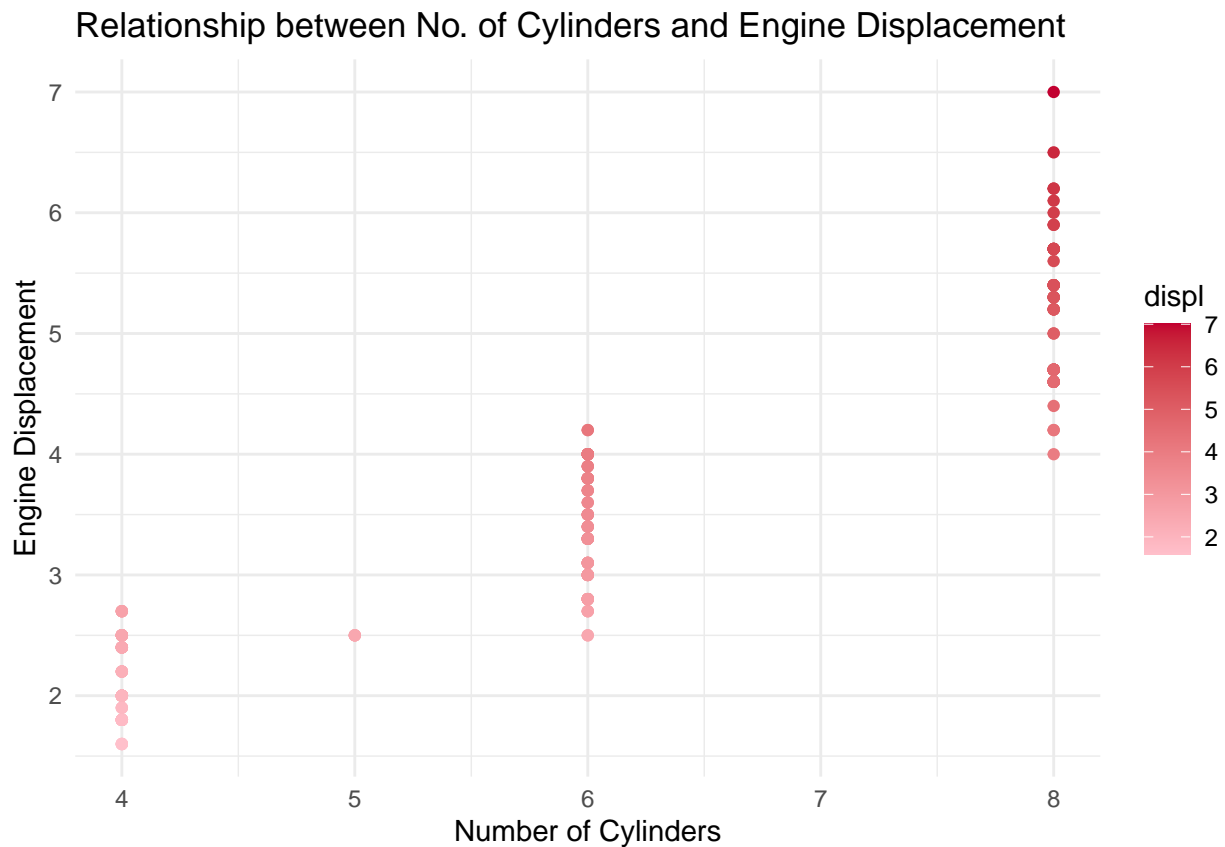
```
ggplot(top_20_data, aes(y = model)) +
  geom_bar(fill = "pink", color = "black") +
  labs(title = "Top 20 Observations: Number of Cars per Model",
       x = "Number of Cars", y = "Model") +
  coord_flip() +
  theme(axis.text.y = element_text(angle = 45, hjust = 1))
```

Top 20 Observations: Number of Cars per Model



#5. Plot the relationship between *cyl* - number of cylinders and *displ* - engine displacement using *geom_point*

```
ggplot(mpg, aes(x = cyl, y = displ, color = displ)) +
  geom_point() +
  labs(title = "Relationship between No. of Cylinders and Engine Displacement",
       x = "Number of Cylinders", y = "Engine Displacement") +
  scale_color_gradient(low = "pink", high = "#C20030") +
  theme_minimal()
```

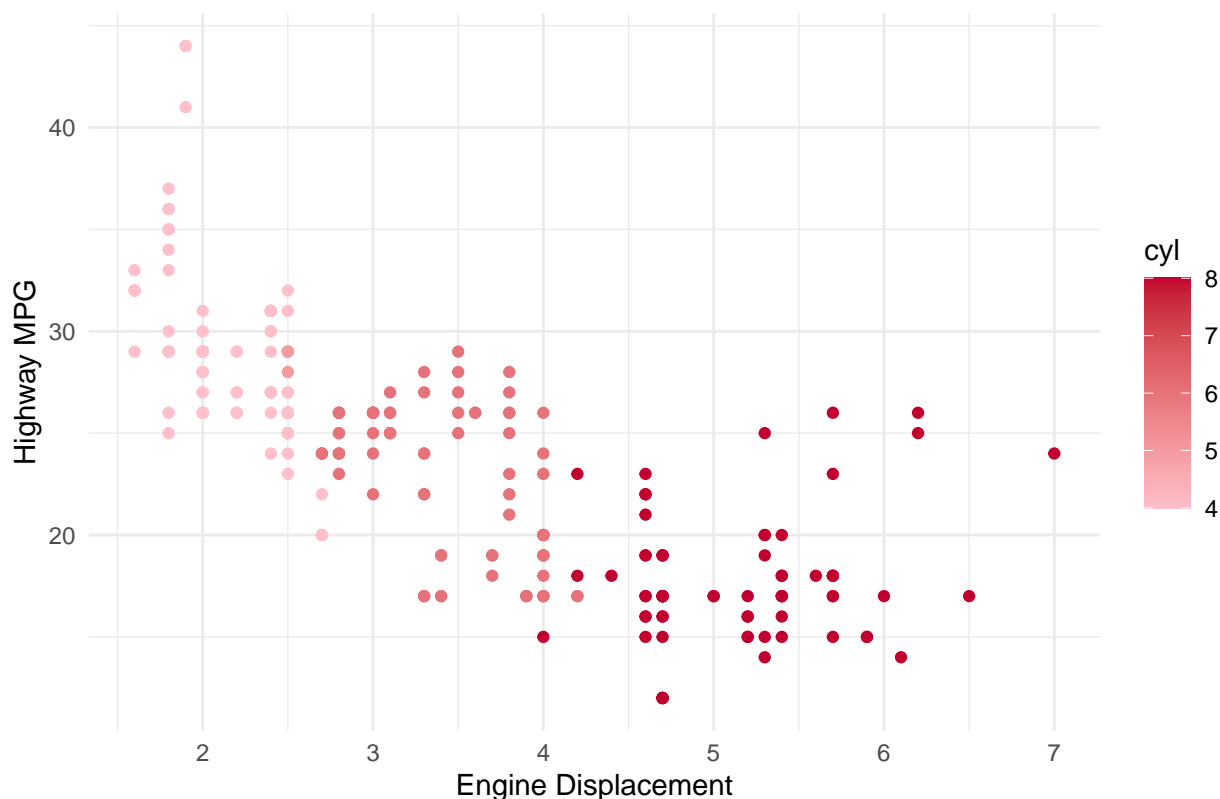


#5a. How would you describe its relationship? Show the codes and its result. The plot shows that there

#5a. How would you describe its relationship? Show the codes and its result.

```
ggplot(mpg, aes(x = displ, y = hwy, color = cyl)) +
  geom_point() +
  labs(title = "Relationship between Engine Displacement and Highway MPG",
       x = "Engine Displacement", y = "Highway MPG") +
  scale_color_gradient(low = "pink", high = "#C20030") +
  theme_minimal()
```

Relationship between Engine Displacement and Highway MPG



#6. Import the traffic.csv onto your R environment.

```
traffic <- read.csv("traffic.csv")
```

#6a. How many numbers of observation does it have? What are the variables of the traffic dataset the Sh

```
trafficdata <- read.csv("traffic.csv")
```

```
number_obs <- nrow(trafficdata)
```

```
var_traffic <- ncol(trafficdata)
```

```
varnames <- names(trafficdata)
```

```
cat("Number of Observations:", number_obs, "\n")
```

```
## Number of Observations: 48120
```

```
cat("Number of Variables:", var_traffic, "\n")
```

```
## Number of Variables: 4
```

```
cat("Variables:", paste(varnames, collapse = ", "), "\n")
```

```
## Variables: DateTime, Junction, Vehicles, ID
```

#6b. subset the traffic dataset into junctions. What is the R codes and its output?

```
junction_traffic <- c(1,2,3,4)
```

```
junction_subset <- trafficdata[trafficdata$Junction %in% junction_traffic, ]
```

#6c. Plot each junction in a using geom_line(). Show your solution and output.

```
library(ggplot2)
```

```
ggplot(trafficdata, aes(x = DateTime, y = Vehicles, color = as.factor(Junction))) +
  geom_line() +
  scale_color_manual(values = c("#ffe5ec", "#ffc2d1", "#ffb3c6", "#ff8fab")) +
  labs(title = "Traffic by Junction",
       x = "Date Time",
       y = "Number of Vehicles") +
  theme_minimal() +
  facet_wrap(~Junction, scales = "free_y", ncol = 1) +
  guides(color = guide_legend(title = "Junction"))
```

`geom_line()`: Each group consists of only one observation.

i Do you need to adjust the group aesthetic?

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i Do you need to adjust the group aesthetic?

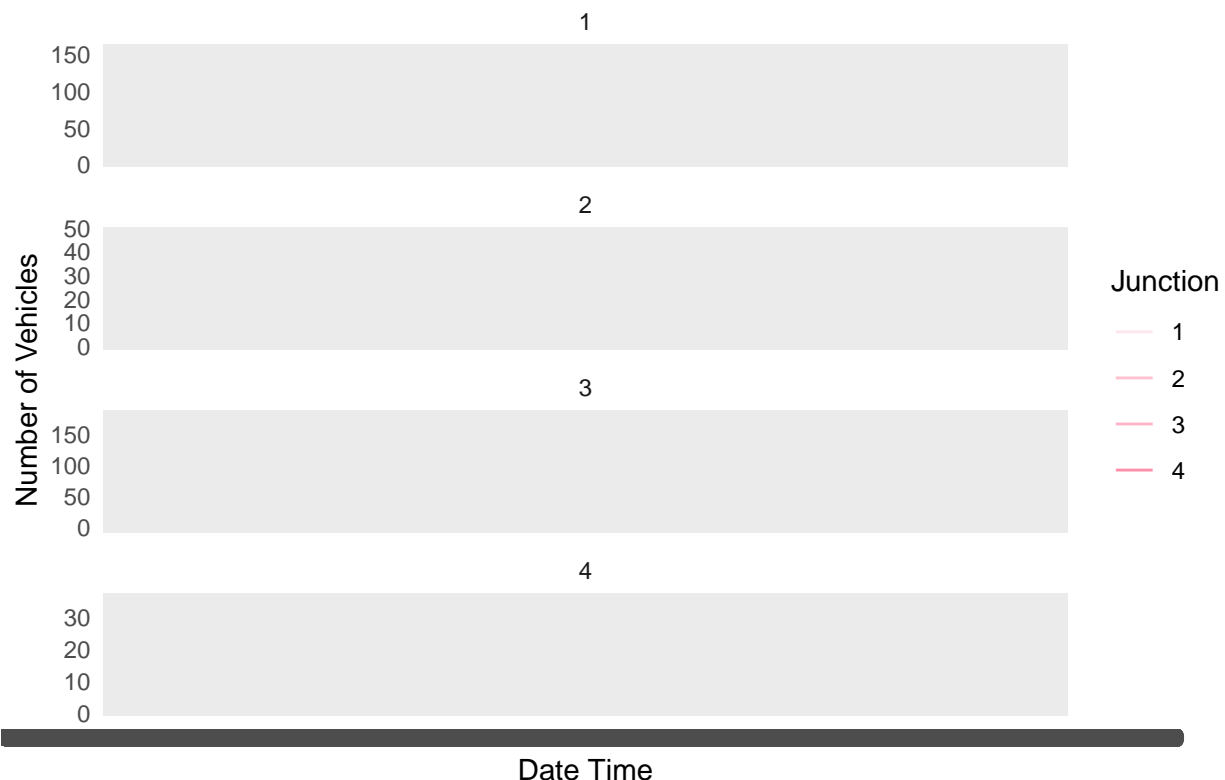
`geom_line()`: Each group consists of only one observation.

i Do you need to adjust the group aesthetic?

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i Do you need to adjust the group aesthetic?

Traffic by Junction



#7. From alexa_file.xlsx, import it to your environment

```
library(readxl)
```

```
alexafile <- read_excel("alexafile.xlsx")
```


#7a. How many observations does alexa_file has? What about the number of columns? Show your solution and answer.

```
alexa_obser <- nrow(alexafile)

alexacol_obser <- ncol(alexafile)

cat("The number of observations on alexa is:", alexa_obser, "\n")
```

```
## The number of observations on alexa is: 3150
```

```
cat("The number of columns on alexa is:", alexacol_obser, "\n")
```

```
## The number of columns on alexa is: 5
```

#7b. group the variations and get the total of each variations. Use dplyr package. Show solution and answer.

```
library(dplyr)
groupvariations <- alexafile %>%
  group_by(variation) %>%
  summarise(totalcount_ = n())
```

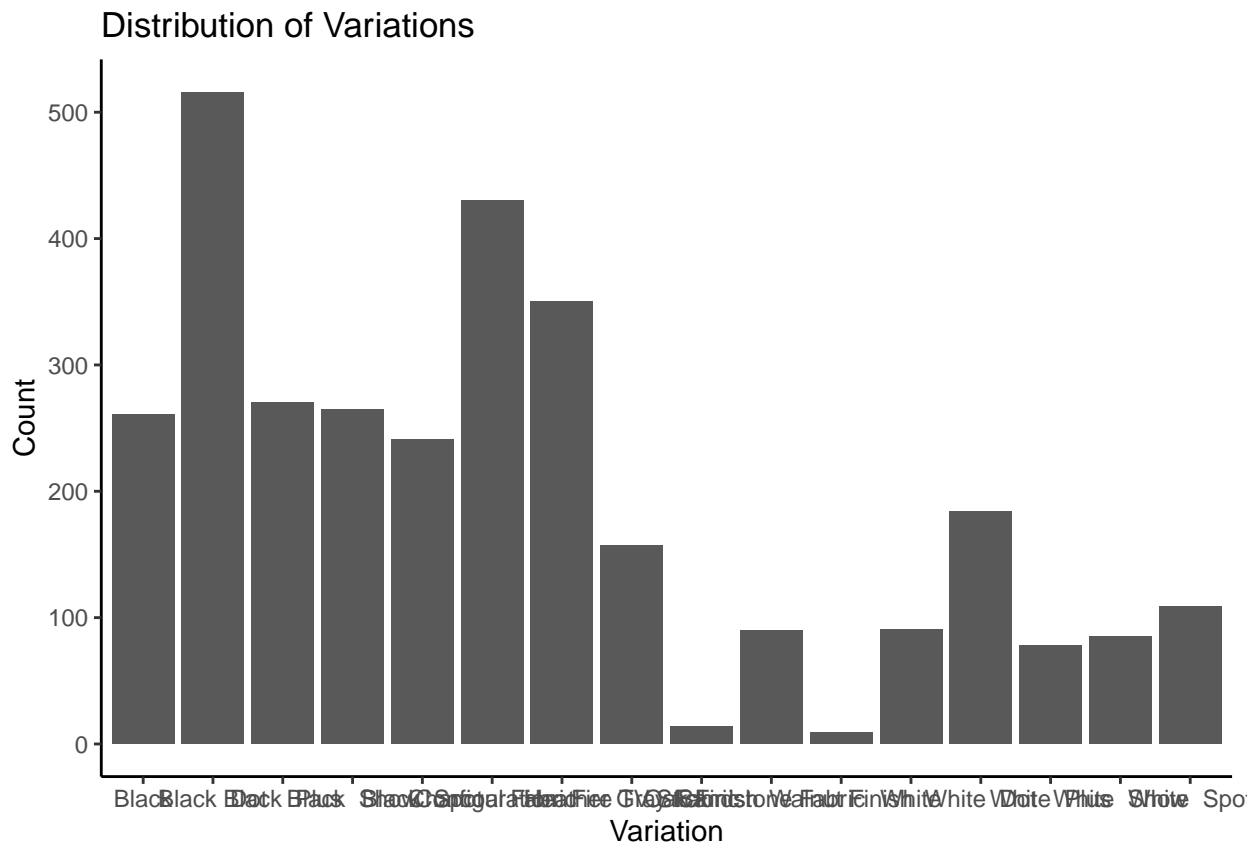
```
groupvariations
```

```
## # A tibble: 16 x 2
##   variation                totalcount_
##   <chr>                  <int>
## 1 Black                    261
## 2 Black Dot                516
## 3 Black Plus               270
## 4 Black Show               265
## 5 Black Spot               241
## 6 Charcoal Fabric          430
## 7 Configuration: Fire TV Stick 350
## 8 Heather Gray Fabric       157
## 9 Oak Finish                14
## 10 Sandstone Fabric          90
## 11 Walnut Finish             9
## 12 White                    91
## 13 White Dot                184
## 14 White Plus               78
## 15 White Show               85
## 16 White Spot              109
```

#7c. Plot the variations using the ggplot() function. What did you observe? Complete the details of the plot.

```
library(ggplot2)

ggplot(alexafile, aes(x = variation)) +
  geom_bar() +
  labs(title = "Distribution of Variations",
       x = "Variation",
       y = "Count") +
  theme_classic()
```



#7d. Plot a `geom_line()` with the date and the number of verified reviews. Complete the details of the g

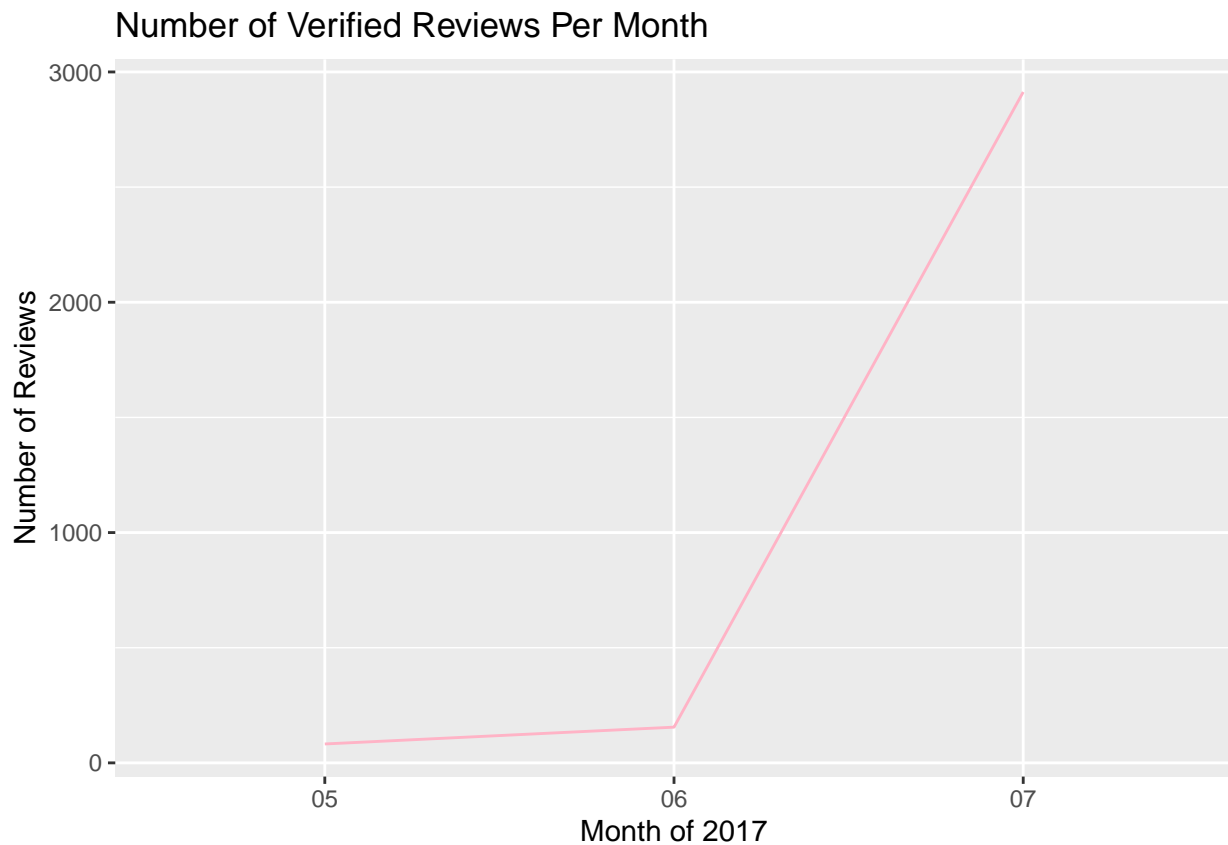
```
alexafile$date <- as.Date(alexafile$date)

alexafile$month <- format(alexafile$date, "%m")

monthscount <- alexafile %>%
  group_by(month) %>%
  summarise(num_reviews = n())

monthlyrev <- table(monthscount)

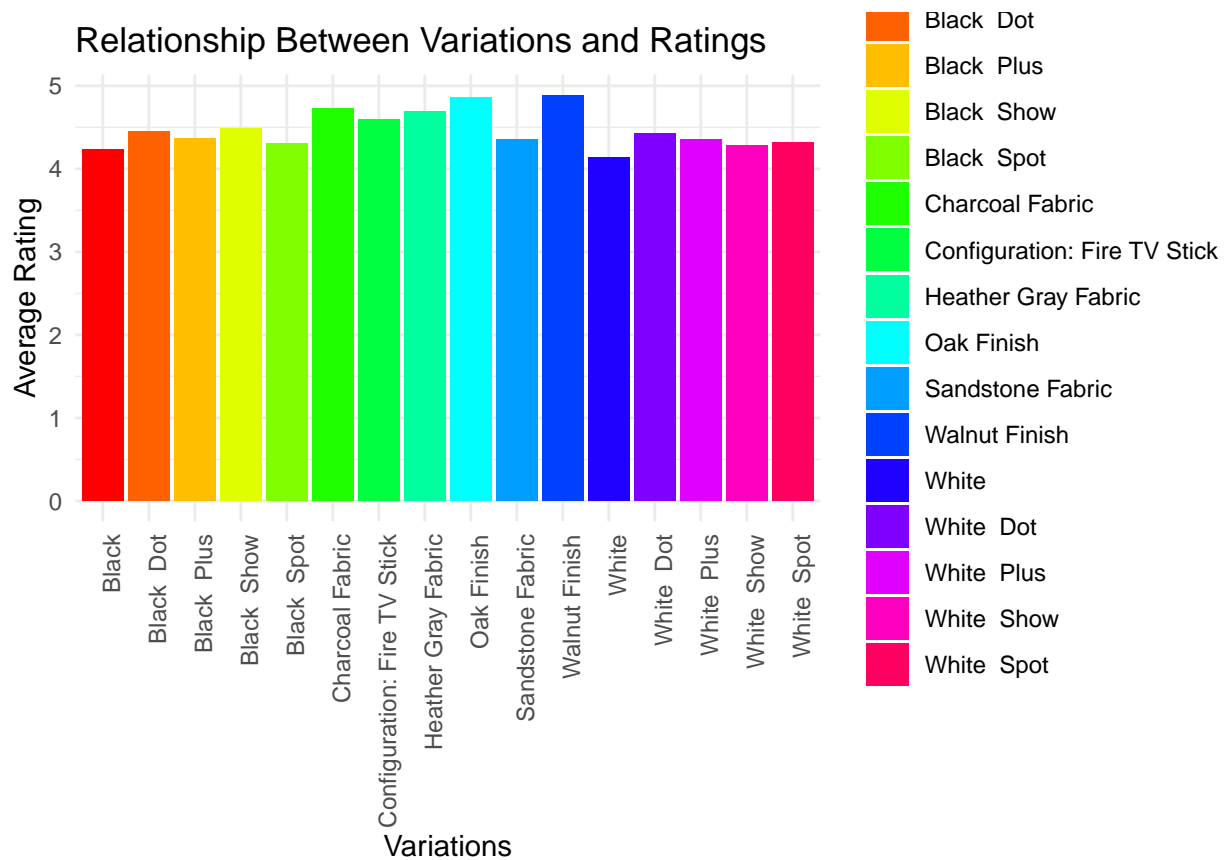
ggplot(monthscount, aes(x = month, y = num_reviews, group = 1)) +
  geom_line(color = "#ffb3c6") +
  labs(title = "Number of Verified Reviews Per Month",
       x = "Month of 2017", y = "Number of Reviews")
```



#7e. Get the relationship of variations and ratings. Which variations got the most highest in rating? Plot the relationship.

```
library(dplyr)

ggplot(alexafile, aes(x = variation, y = rating, fill = variation)) +
  geom_bar(stat = "summary", fun = "mean", position = "dodge") +
  scale_fill_manual(values = rainbow(n = length(unique(alexafile$variation)))) +
  labs(title = "Relationship Between Variations and Ratings",
       x = "Variations",
       y = "Average Rating") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 90, hjust = 1))
```



```
var_ratings <- alexafile %>%
  group_by(variation)%>%
  summarise(average_rating = mean(rating, na.rm = TRUE))

max_rating <- max(var_ratings$average_rating, na.rm = TRUE)
```

```
## Warning: Unknown or uninitialised column: `average_rating`.
## Warning in max(var_ratings$average_rating, na.rm = TRUE): no non-missing
## arguments to max; returning -Inf
```

```
highrate <- alexafile %>%
  filter(rating == max_rating)
print(highrate)
```

```
## # A tibble: 0 x 6
## # i 6 variables: rating <dbl>, date <date>, variation <chr>,
## #   verified_reviews <chr>, feedback <dbl>, month <chr>
```