

# RWorksheet\_calvario#4c.Rmd

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2024-11-19

## 1. Use the dataset mpg

### a. Solutions on how to import a csv file into the environment.

```
library(ggplot2)
```

```
mpg_data <- read.csv("mpg.csv")  
str(mpg_data)
```

```
## 'data.frame':   234 obs. of  12 variables:  
## $ X           : int  1 2 3 4 5 6 7 8 9 10 ...  
## $ manufacturer: chr  "audi" "audi" "audi" "audi" ...  
## $ model       : chr  "a4" "a4" "a4" "a4" ...  
## $ displ       : num  1.8 1.8 2 2 2.8 2.8 3.1 1.8 1.8 2 ...  
## $ year        : int  1999 1999 2008 2008 1999 1999 2008 1999 1999 2008 ...  
## $ cyl         : int  4 4 4 4 6 6 6 4 4 4 ...  
## $ trans       : chr  "auto(l5)" "manual(m5)" "manual(m6)" "auto(av)" ...  
## $ drv         : chr  "f" "f" "f" "f" ...  
## $ cty         : int  18 21 20 21 16 18 18 18 16 20 ...  
## $ hwy         : int  29 29 31 30 26 26 27 26 25 28 ...  
## $ fl         : chr  "p" "p" "p" "p" ...  
## $ class       : chr  "compact" "compact" "compact" "compact" ...
```

b. Which variables from mpg dataset are categorical?

The categorical variables in the mpg dataset are manufacturer, model, trans, drv, fl, and class. These represent distinct groups or categories, such as car brand, transmission type, drivetrain, fuel type, and car class.

c. Which are continuous variables?

The continuous variables in the mpg dataset are displ, cty, hwy, and cyl, as they represent measurable numerical values.

2.1. Which manufacturer has the most models in this data set? Which model has the most variations? Show your answer.

a. 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_model <- mpg %>%
  group_by(manufacturer) %>%
  summarize(model_num = n_distinct(model)) %>%
  arrange(desc(model_num))

manufacturer_model

## # A tibble: 15 x 2
##   manufacturer model_num
##   <chr>          <int>
## 1 toyota          6
## 2 chevrolet       4
## 3 dodge           4
## 4 ford            4
## 5 volkswagen      4
## 6 audi            3
## 7 nissan           3
## 8 hyundai         2
## 9 subaru          2
## 10 honda          1
```

```
## 11 jeep 1
## 12 land rover 1
## 13 lincoln 1
## 14 mercury 1
## 15 pontiac 1

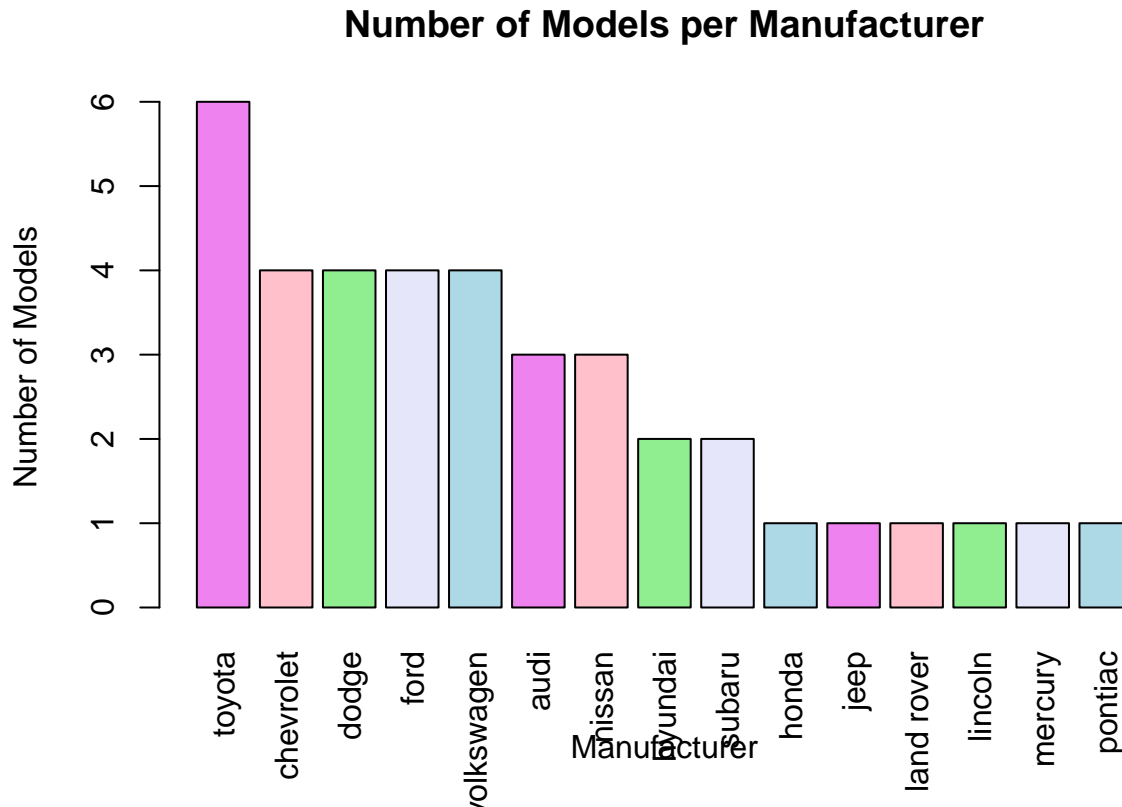
variations_num <- table(mpg$model)
variations_num [variations_num == max(variations_num)]

## caravan 2wd
## 11
```

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

```
manufacturer_data <- setNames(
  manufacturer_model$model_num,
  manufacturer_model$manufacturer
)

barplot(manufacturer_data,
  main = "Number of Models per Manufacturer",
  xlab = "Manufacturer",
  ylab = "Number of Models",
  col = c("violet", "pink", "lightgreen", "lavender", "lightblue"),
  las = 3)
```



```

variations_num <- mpg %>%
  group_by(model) %>%
  summarize(count = n()) %>%
  arrange(desc(count))

```

```
variations_num
```

```

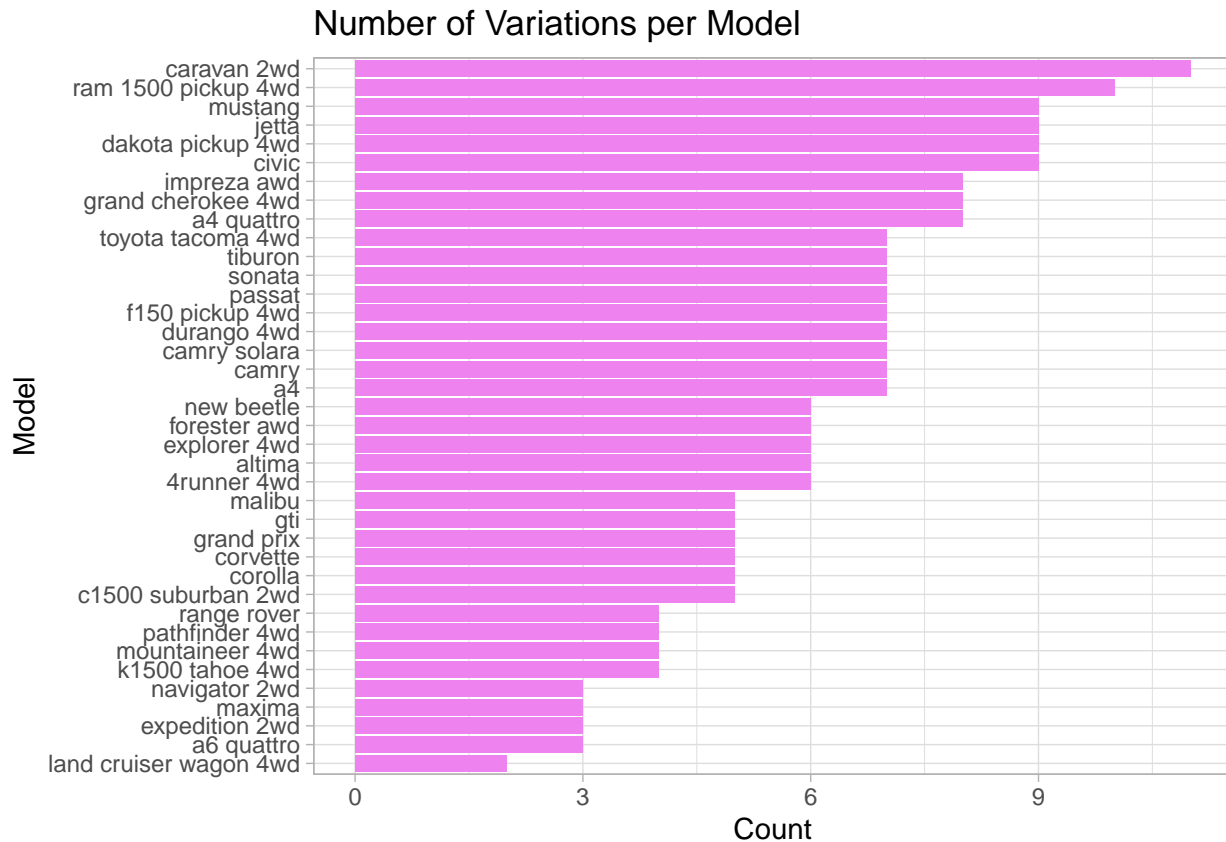
## # A tibble: 38 x 2
##   model          count
##   <chr>         <int>
## 1 caravan 2wd         11
## 2 ram 1500 pickup 4wd  10
## 3 civic              9
## 4 dakota pickup 4wd    9
## 5 jetta              9
## 6 mustang             9
## 7 a4 quattro          8
## 8 grand cherokee 4wd   8
## 9 impreza awd         8
## 10 a4                 7
## # i 28 more rows

```

```

ggplot(variations_num, aes(x = reorder(model, count), y = count)) +
  geom_bar(stat = "identity", fill = "violet") +
  coord_flip() +
  labs(title = "Number of Variations per Model", x = "Model", y = "Count") +
  theme_light()

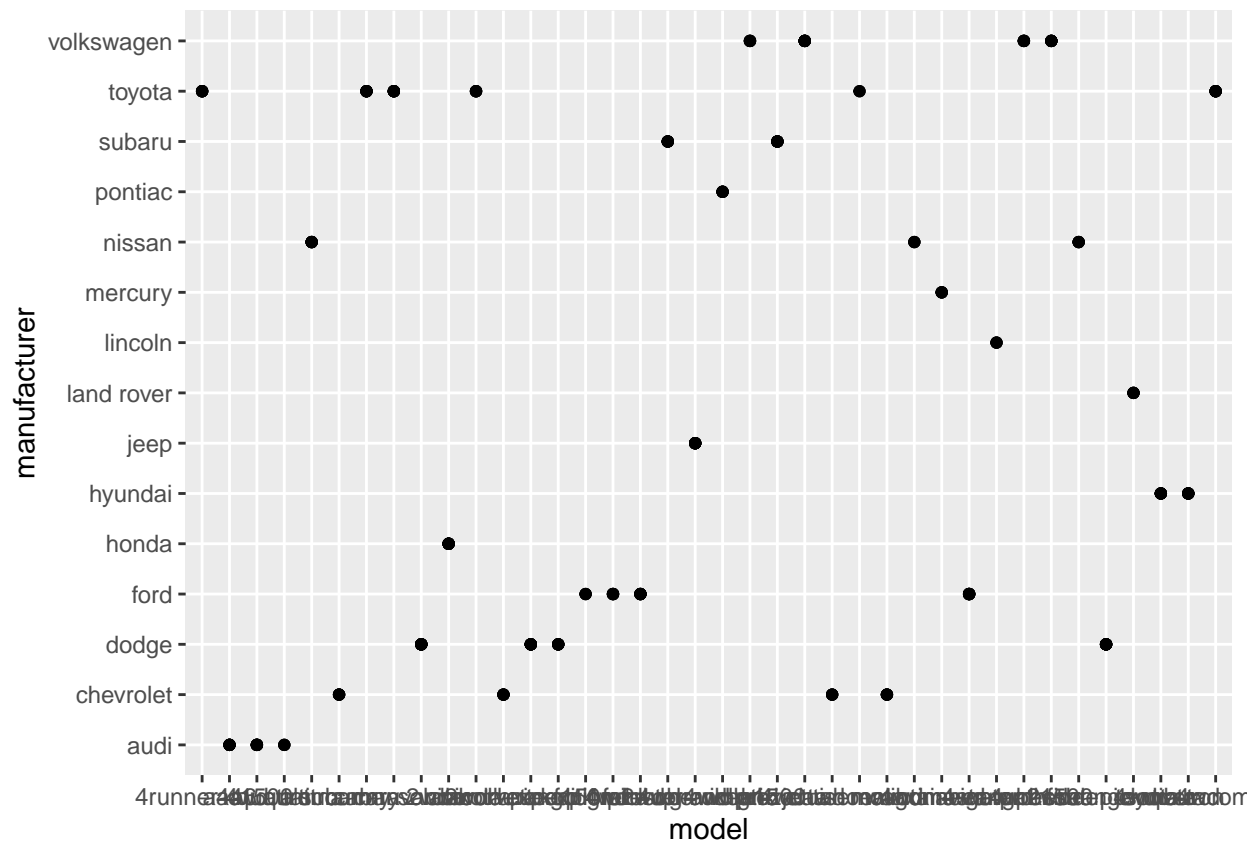
```



**2.2.** Same dataset will be used. You are going to show the relationship of the model and the manufacturer.

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

```
ggplot(mpg, aes(model, manufacturer)) + geom_point()
```



`ggplot(mpg, aes(model, manufacturer)) + geom_point()` creates a scatter plot showing the relationship between car models (model) on the x-axis and manufacturers (manufacturer) on the y-axis, with each point representing a car observation.

b. For you, is it useful? If not, how could you modify the data to make it more informative?

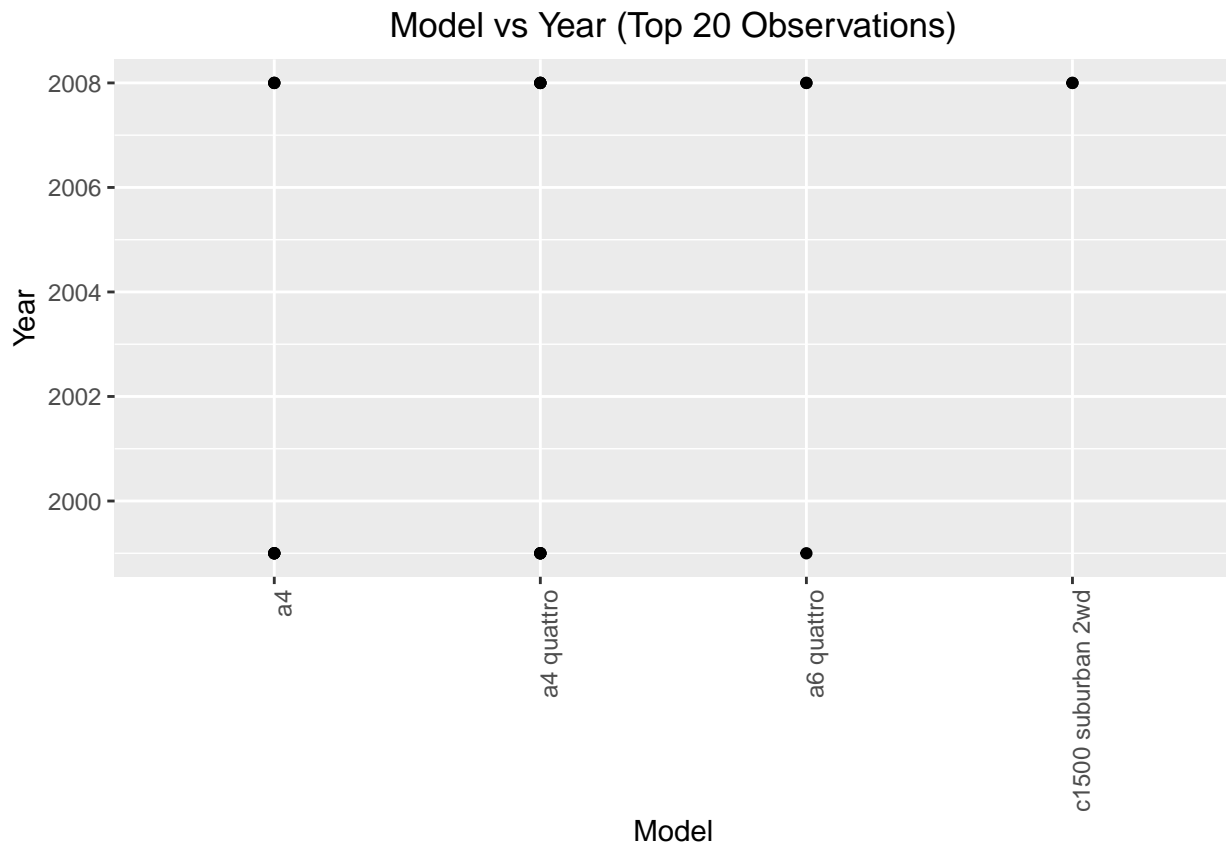
In its current form, this plot isn't very useful as it doesn't effectively visualize the relationship between car models and manufacturers. Since both variables are categorical, a scatter plot isn't the most appropriate way to represent this relationship.

3. Plot the model and the year using `ggplot()`. Use only the top 20 observations. Write the codes and its results.

```
library(ggplot2)

# Top 20 observations from the mpg_data dataset
top_20_data <- head(mpg_data, 20)

# Plot using ggplot2
ggplot(top_20_data, aes(x = model, y = year)) +
  geom_point() +
  labs(
    title = "Model vs Year (Top 20 Observations)",
    x = "Model",
    y = "Year"
  ) +
  theme(
    axis.text.x = element_text(angle = 90, hjust = 1),
    plot.title = element_text(hjust = 0.5) # Centering the title
  )
```



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

a. Plot using `geom_bar()` using the top 20 observations only. The graphs should have a title, labels and colors. Show code and results.

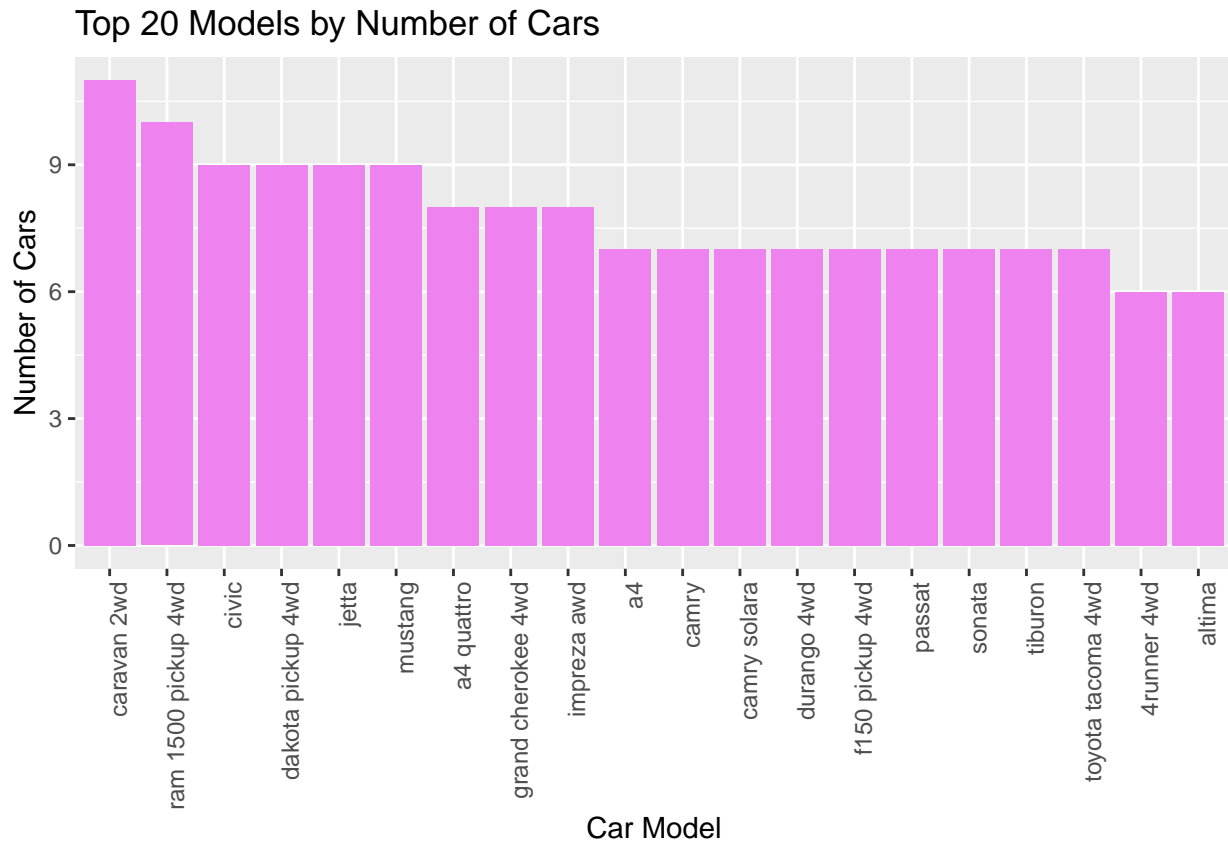
```
library(dplyr)
library(ggplot2)

car_counts <- mpg_data %>%
  group_by(model) %>%
  summarise(count = n())

top_20_models <- car_counts %>%
  arrange(desc(count)) %>%
  head(20)

ggplot(top_20_models, aes(x = reorder(model, -count), y = count)) +
  geom_bar(stat = "identity", fill = "violet") +
  labs(
    title = "Top 20 Models by Number of Cars",
    x = "Car Model",
    y = "Number of Cars"
  ) +
```

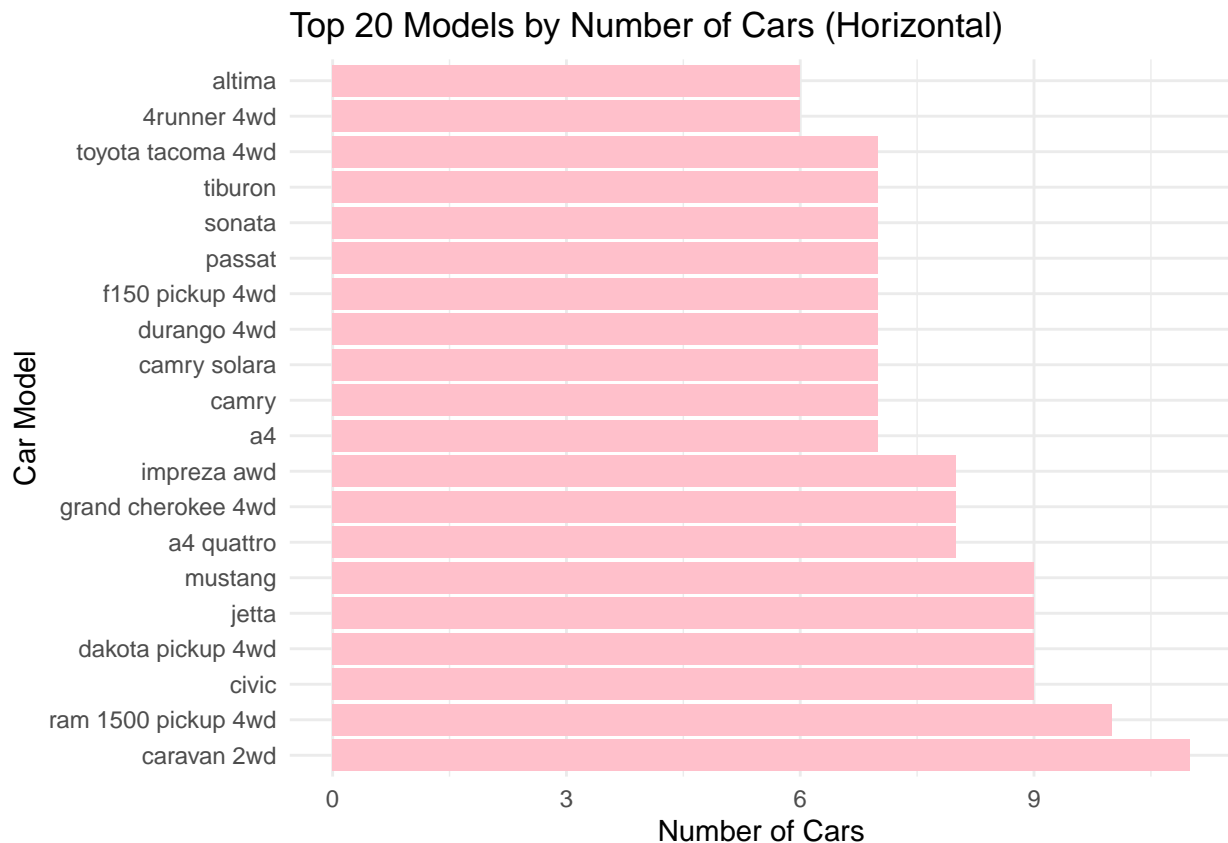
```
theme(axis.text.x = element_text(angle = 90, hjust = 1)) # Rotate x-axis labels
```



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

```
ggplot(top_20_models, aes(x = reorder(model, -count), y = count)) +
  geom_bar(stat = "identity", fill = "pink") +
  labs(
    title = "Top 20 Models by Number of Cars (Horizontal)",
    x = "Car Model",
    y = "Number of Cars"
  ) +
  coord_flip() +
  theme_minimal()
```



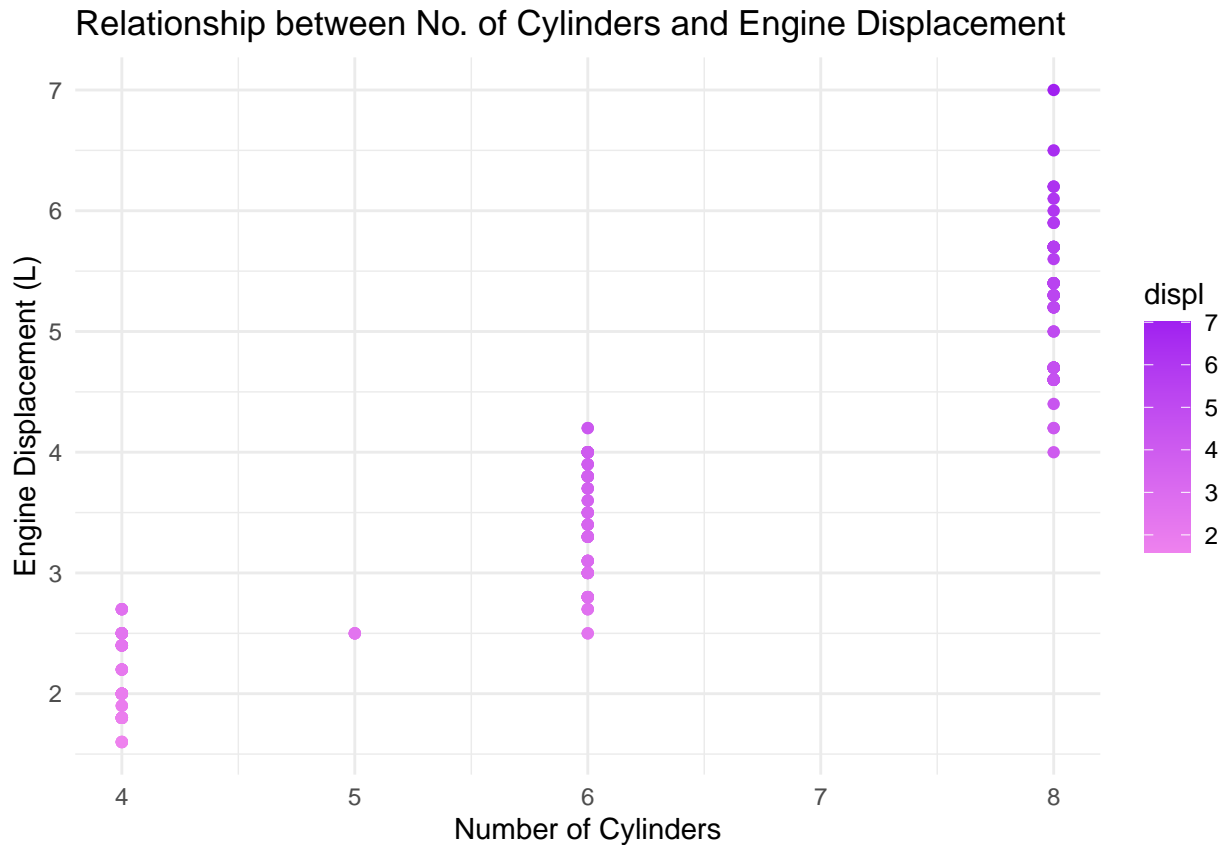


5. Plot the relationship between cyl - number of cylinders and displ - engine displacement using `geom_point` with aesthetic color = engine displacement. Title should be “Relationship between No. of Cylinders and Engine Displacement”.

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

```
library(ggplot2)

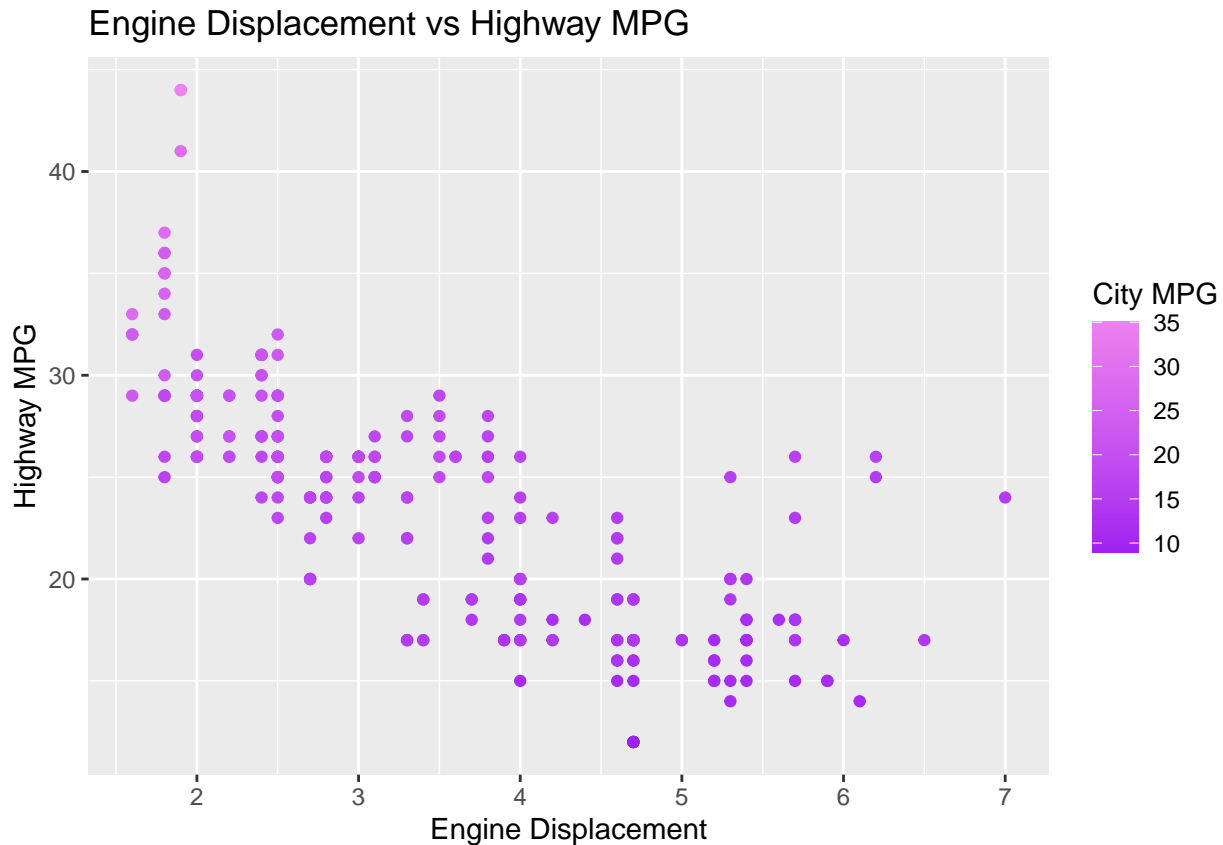
ggplot(mpg_data, aes(x = cyl, y = displ, color = displ)) +
  geom_point() +
  labs(
    title = "Relationship between No. of Cylinders and Engine Displacement",
    x = "Number of Cylinders",
    y = "Engine Displacement (L)"
  ) +
  scale_color_gradient(low = "violet", high = "purple") +
  theme_minimal()
```



The plot will show a scatter plot with the number of cylinders on the x-axis and the engine displacement on the y-axis. The points will be colored based on the engine displacement, where lower displacements will be shaded in violet, and higher displacements will be shaded in purple.

6. Plot the relationship between `displ` (engine displacement) and `hwy` (highway miles per gallon). Mapped it with a continuous variable you have identified in #1-c.

```
ggplot(mpg_data, aes(x = displ, y = hwy, color = cty)) +
  geom_point() +
  labs(title = "Engine Displacement vs Highway MPG",
        x = "Engine Displacement", y = "Highway MPG", color = "City MPG") +
  scale_color_gradient(low = "purple", high = "violet")
```



What is its result?

The results of this plot help explain how engine displacement and vehicle weight both affect fuel efficiency, revealing the trade-offs between engine size, vehicle weight, and fuel economy.

Why it produced such output?

The plot shows that as engine displacement (displ) increases, highway miles per gallon (hwy) decreases, with heavier vehicles (mapped by weight) generally having larger engines and lower fuel efficiency.

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

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

```
##           DateTime Junction Vehicles          ID
## 1 2015-11-01 00:00:00         1      15 20151101001
## 2 2015-11-01 01:00:00         1      13 20151101011
```

```
## 3 2015-11-01 02:00:00      1      10 20151101021
## 4 2015-11-01 03:00:00      1       7 20151101031
## 5 2015-11-01 04:00:00      1       9 20151101041
## 6 2015-11-01 05:00:00      1       6 20151101051
```

a. How many numbers of observation does it have? What are the variables of the traffic dataset? Show your answer.

```
n_obs <- nrow(traffic)

variables <- colnames(traffic)

n_obs
```

```
## [1] 48120
```

```
variables
```

```
## [1] "DateTime" "Junction" "Vehicles" "ID"
```

There are 48120 observations. The variables in the traffic dataset are: “DateTime” “Junction” “Vehicles” “ID”.

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

```
library(dplyr)

junction_data <- traffic %>%
  group_by(Junction) %>%
  group_split()

print(junction_data[[1]])
```

```
## # A tibble: 14,592 x 4
##   DateTime      Junction Vehicles      ID
##   <chr>          <int>    <int>    <dbl>
## 1 2015-11-01 00:00:00      1      15 20151101001
## 2 2015-11-01 01:00:00      1      13 20151101011
## 3 2015-11-01 02:00:00      1      10 20151101021
## 4 2015-11-01 03:00:00      1       7 20151101031
## 5 2015-11-01 04:00:00      1       9 20151101041
## 6 2015-11-01 05:00:00      1       6 20151101051
## 7 2015-11-01 06:00:00      1       9 20151101061
## 8 2015-11-01 07:00:00      1       8 20151101071
## 9 2015-11-01 08:00:00      1      11 20151101081
## 10 2015-11-01 09:00:00      1      12 20151101091
## # i 14,582 more rows
```

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

```
library(ggplot2)

traffic$DateTime <- as.POSIXct(traffic$DateTime, format = "%Y-%m-%d %H:%M:%S")

ggplot(traffic, aes(x = DateTime, y = Vehicles, color = factor(Junction))) +
  geom_line() + # Create the line plot
  labs(title = "Traffic Counts by Junction Over Time",
       x = "Date and Time",
       y = "Number of Vehicles",
       color = "Junction") +
  scale_color_manual(values = c("1" = "#8A2BE2", # Violet
                                "2" = "#9400D3", # Dark Violet
                                "3" = "#EE82EE", # Light Violet
                                "4" = "#DA70D6", # Orchid (light violet shade)
                                "5" = "#E6E6FA")) +

  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) # Rotate x-axis labels for better readability
```

