

# RWorksheet\_Soldevilla#4c.Rmd

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```
install.packages("ggplot2")

## Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.3'
## (as 'lib' is unspecified)

install.packages("dplyr")

## Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.3'
## (as 'lib' is unspecified)

install.packages("tidyverse")

## Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.3'
## (as 'lib' is unspecified)

#1

#1a

# Show your solutions on how to import a csv file into the environment
library(readr)
mpg <- read_csv("mpg.csv")

## New names:
## Rows: 234 Columns: 12
## -- Column specification
## ----- Delimiter: "," chr
## (6): manufacturer, model, trans, drv, fl, class dbl (6): ...1, displ, year,
## cyl, cty, hwy
## i Use `spec()` to retrieve the full column specification for this data. i
## Specify the column types or set `show_col_types = FALSE` to quiet this message.
## * `` -> `...1`

mpg

## # A tibble: 234 x 12
##   ...1 manufacturer model      displ  year  cyl trans  drv      cty   hwy fl      <dbl> <dbl> <dbl> <chr> <chr> <dbl> <dbl> <chr>
##   <dbl> <chr>      <chr>      <dbl> <dbl> <dbl> <chr> <chr> <dbl> <dbl> <chr>
## 1     1 audi      a4         1.8  1999    4 auto~ f      18   29 p
## 2     2 audi      a4         1.8  1999    4 manu~ f      21   29 p
## 3     3 audi      a4         2    2008    4 manu~ f      20   31 p
## 4     4 audi      a4         2    2008    4 auto~ f      21   30 p
## 5     5 audi      a4         2.8  1999    6 auto~ f      16   26 p
## 6     6 audi      a4         2.8  1999    6 manu~ f      18   26 p
## 7     7 audi      a4         3.1  2008    6 auto~ f      18   27 p
## 8     8 audi      a4 quattro  1.8  1999    4 manu~ 4      18   26 p
## 9     9 audi      a4 quattro  1.8  1999    4 auto~ 4      16   25 p
```

```
## 10      10 audi          a4 quattro    2      2008      4 manu~ 4          20      28 p
## # i 224 more rows
## # i 1 more variable: class <chr>

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

#The variables that are categorical in mpg dataset are manufacturer, model, trans, drv, and fl.

#1c
#Which are continuous variables?

#The continuous variables in the mpg dataset are displ, year, cyl, cty, and hwy.

#2
#Which manufacturer has the most models in this data set? Which model has the most variations? 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_most_models <- mpg %>%
  group_by(manufacturer) %>%
  summarize(number_of_models = n_distinct(model)) %>%
  top_n(1, number_of_models)
model_most_variations <- mpg %>%
  group_by(model) %>%
  summarize(number_of_variations = n_distinct(trans)) %>%
  top_n(1, number_of_variations)
cat("Manufacturer with the most models:", manufacturer_most_models$manufacturer, "\n")

## Manufacturer with the most models: toyota

cat("Model with the most variations:", model_most_variations$model, "\n")

## Model with the most variations: a4 a4 quattro altima camry civic dakota pickup 4wd explorer 4wd gti

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

library(dplyr)
unique_models_by_manufacturer <- mpg %>%
  group_by(manufacturer) %>%
  distinct(model)
print(unique_models_by_manufacturer)

## # A tibble: 38 x 2
## # Groups:   manufacturer [15]
```

```
##      manufacturer model
##      <chr>          <chr>
## 1 audi             a4
## 2 audi             a4 quattro
## 3 audi             a6 quattro
## 4 chevrolet        c1500 suburban 2wd
## 5 chevrolet        corvette
## 6 chevrolet        k1500 tahoe 4wd
## 7 chevrolet        malibu
## 8 dodge            caravan 2wd
## 9 dodge            dakota pickup 4wd
## 10 dodge           durango 4wd
## # i 28 more rows
```

#2b

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

```
models_per_manufacturer <- table(mpg$manufacturer)
```

*#Using plot()*

```
plot(models_per_manufacturer,
      main = "Number of Models by Manufacturer",
      xlab = "Manufacturer",
      ylab = "Number of Models",
      col = "yellow",
      ylim = c(0, max(models_per_manufacturer) + 2))
```

```
max_manufacturer <- which.max(models_per_manufacturer)
text(max_manufacturer, models_per_manufacturer[max_manufacturer] + 1,
     labels = paste("Max:", max(models_per_manufacturer)),
     col = "blue", pos = 3)
```

```
library(ggplot2)
```

```
##
```

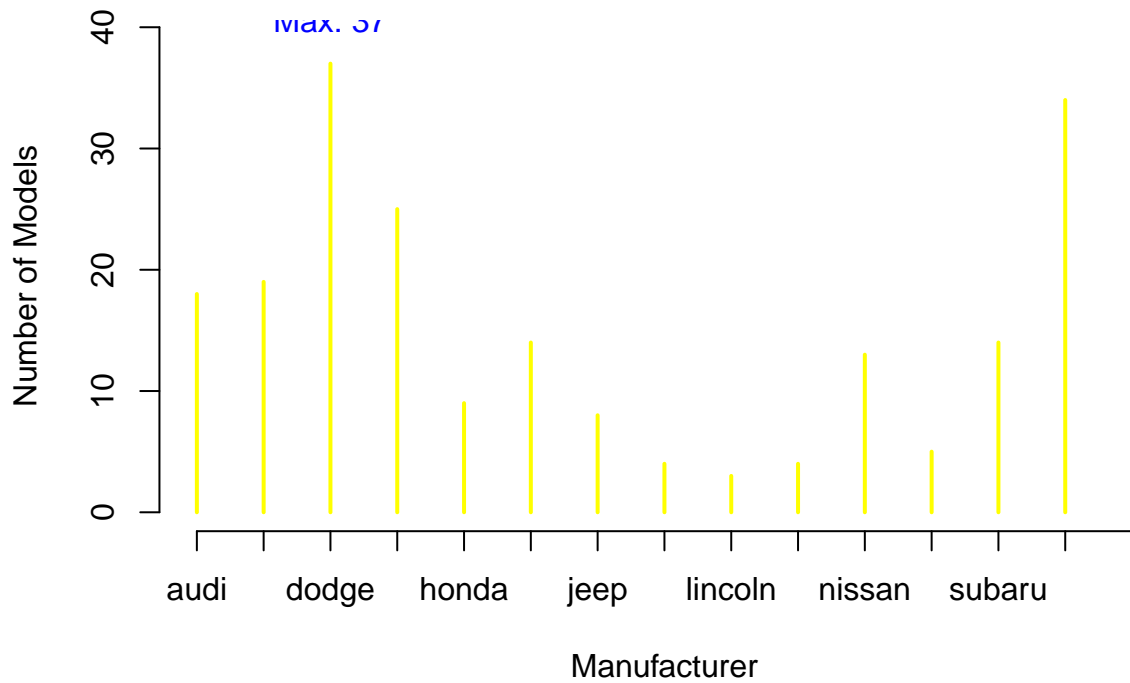
```
## Attaching package: 'ggplot2'
```

```
## The following object is masked _by_ '.GlobalEnv':
```

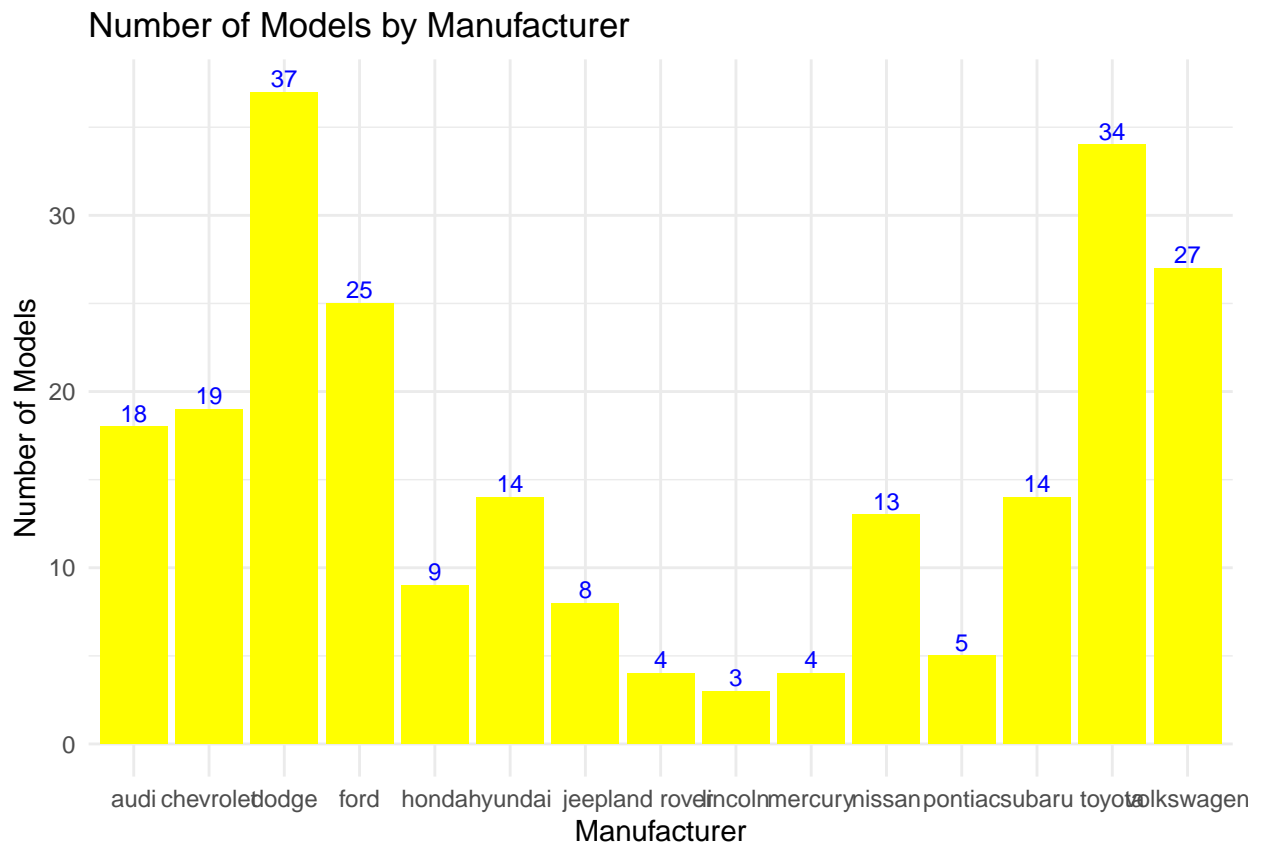
```
##
```

```
##      mpg
```

## Number of Models by Manufacturer



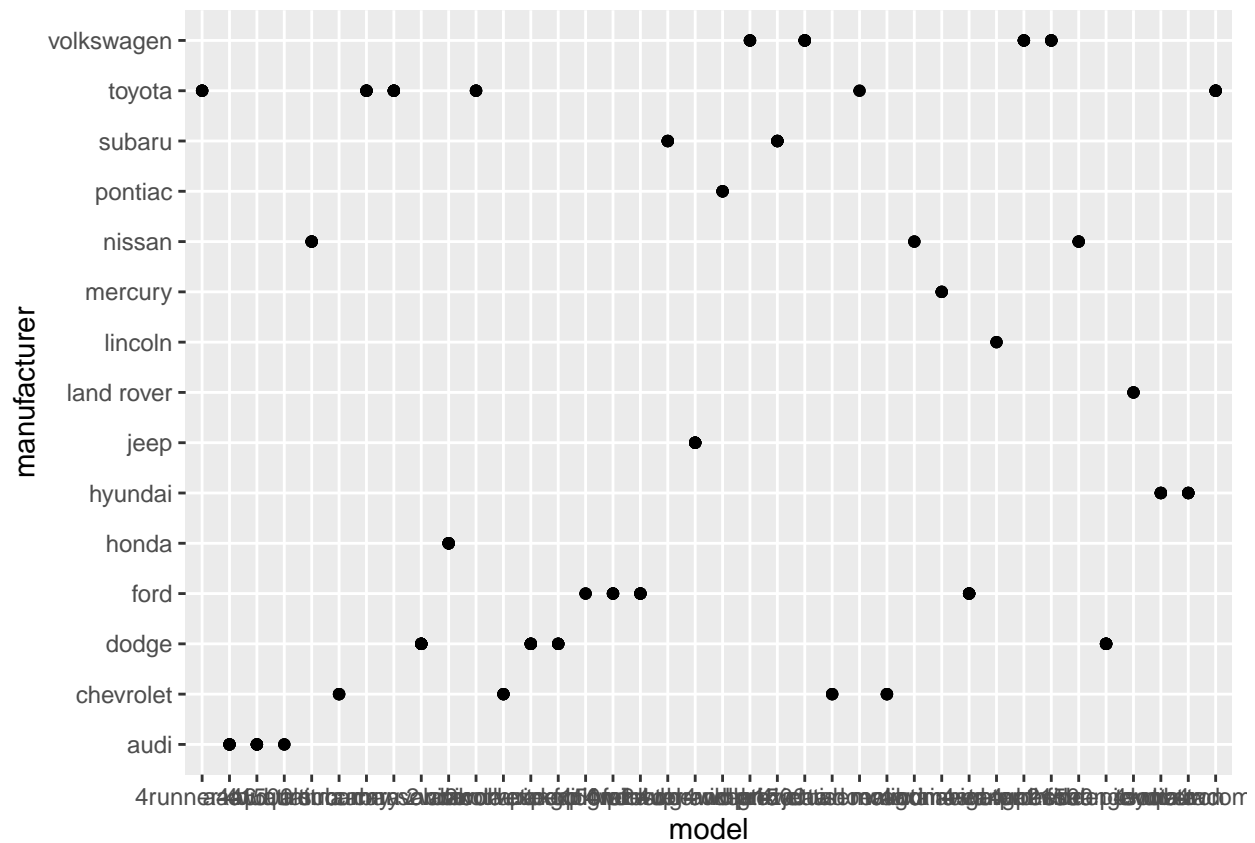
```
#Using ggplot2
ggplot(data = as.data.frame(models_per_manufacturer), aes(x = Var1, y = Freq)) +
  geom_bar(stat = "identity", fill = "yellow") +
  labs(title = "Number of Models by Manufacturer",
       x = "Manufacturer",
       y = "Number of Models") +
  geom_text(aes(label = Freq), vjust = -0.3, col = "blue", size = 3) +
  theme_minimal()
```



#2aa

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

```
library(ggplot2)
ggplot(mpg, aes(model, manufacturer)) +
  geom_point()
```

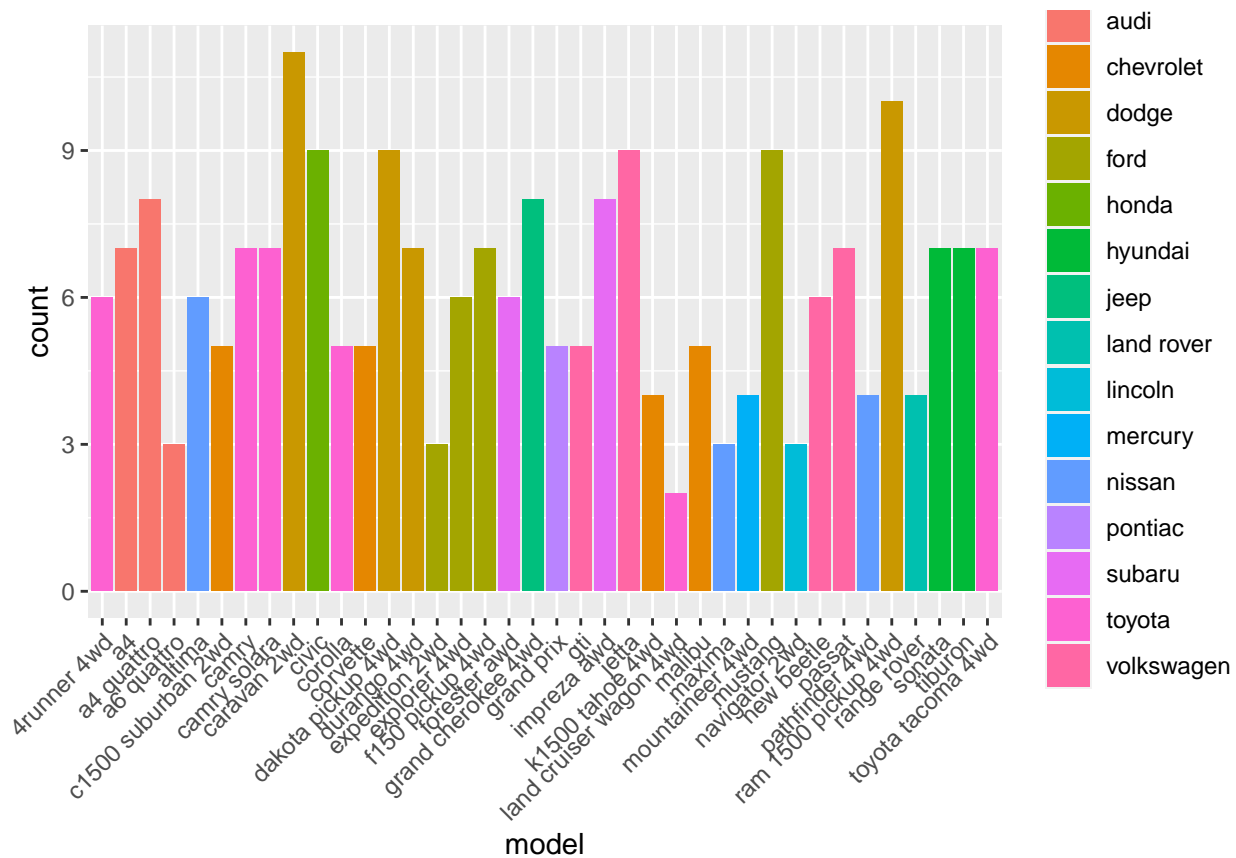


#2bb

# The plot might not be very informative because it's attempting to create a scatter plot using categor

#Tallying Data

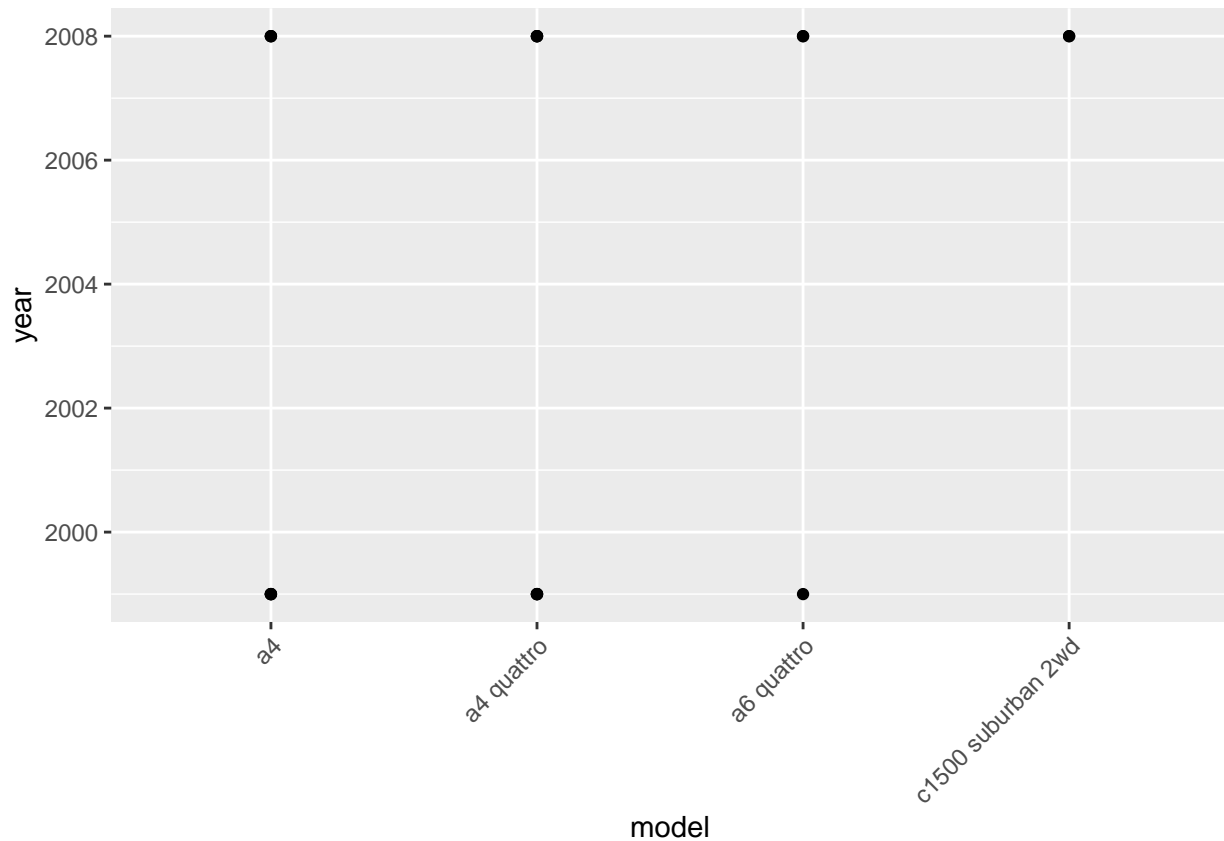
```
library(ggplot2)
ggplot(mpg, aes(model, fill = manufacturer)) +
  geom_bar(position = "dodge") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



#3

*#Plot the model and the year using ggplot(). Use only the top 20 observations. Write the codes and its*

```
library(ggplot2)
top_20 <- head(mpg, 20)
ggplot(top_20, aes(model, year)) +
  geom_point() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



#4

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

```
library(dplyr)
car_counts <- mpg %>%
  group_by(model) %>%
  summarize(number_of_cars = n())
print(car_counts)
```

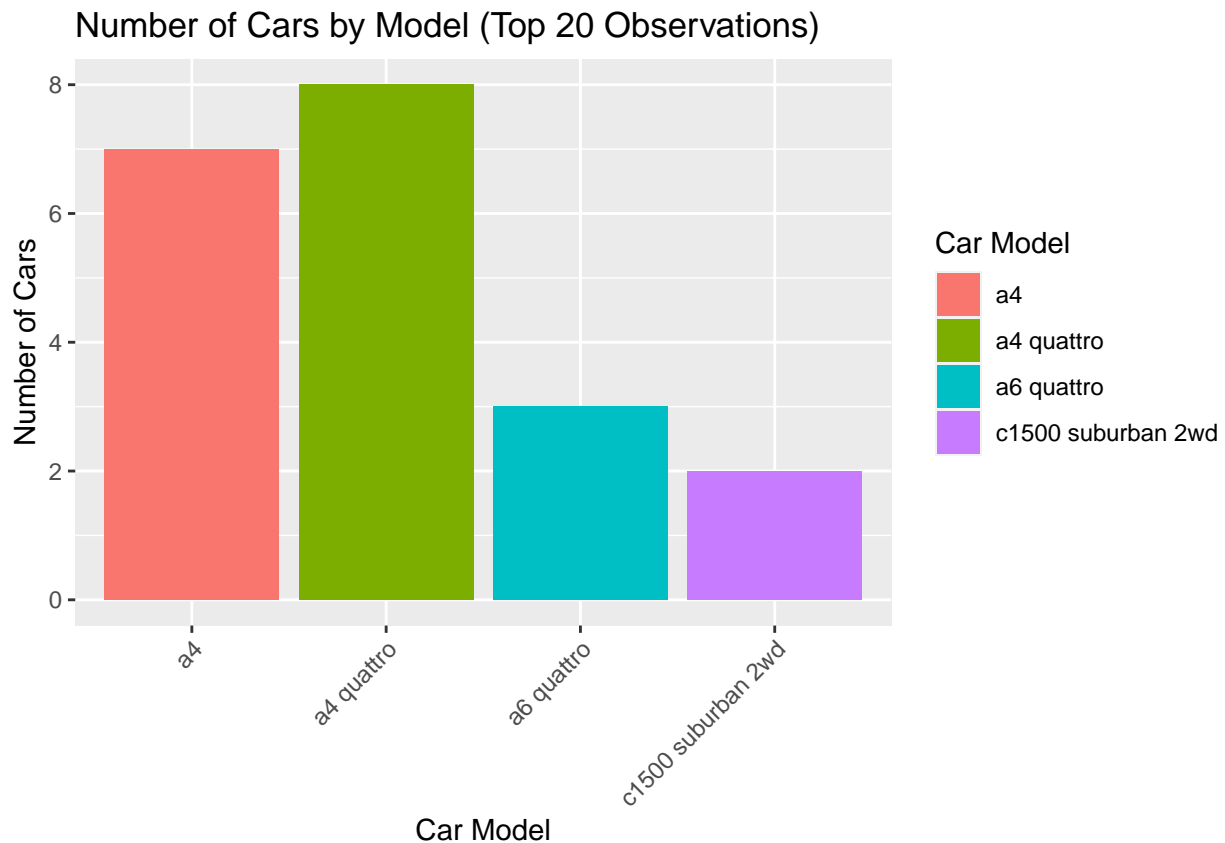
```
## # A tibble: 38 x 2
##   model          number_of_cars
##   <chr>              <int>
## 1 4runner 4wd             6
## 2 a4                     7
## 3 a4 quattro             8
## 4 a6 quattro             3
## 5 altima                 6
## 6 c1500 suburban 2wd     5
## 7 camry                  7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

#4a

*# a. Plot using geom\_bar() using the top 20 observations only. The graphs should have a title, labels a*



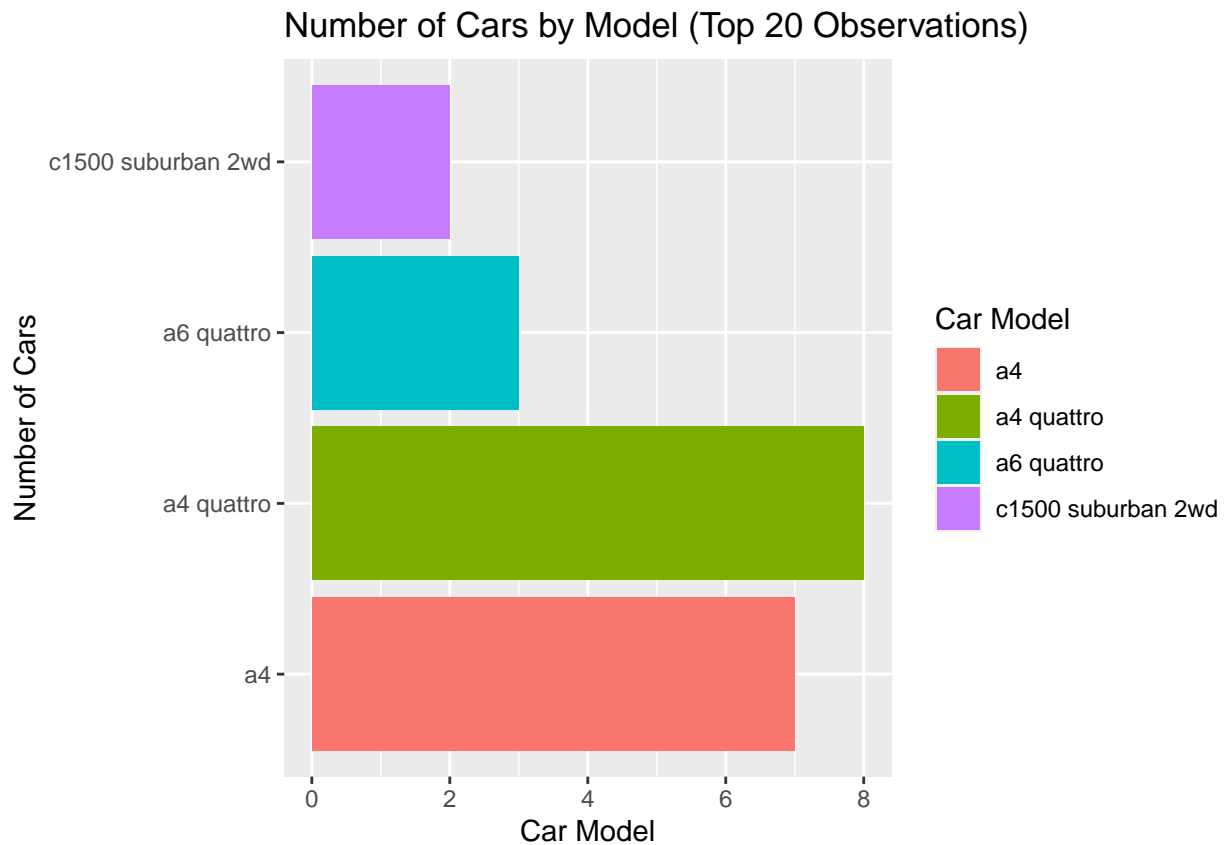
```
library(ggplot2)
top_20 <- head(mpg, 20)
ggplot(top_20, aes(x = model, fill = factor(model))) +
  geom_bar() +
  labs(title = "Number of Cars by Model (Top 20 Observations)",
       x = "Car Model",
       y = "Number of Cars") +
  scale_fill_discrete(name = "Car Model") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



#4b

*# Plot using the geom\_bar() + coord\_flip() just like what is shown below. Show codes and its result.*

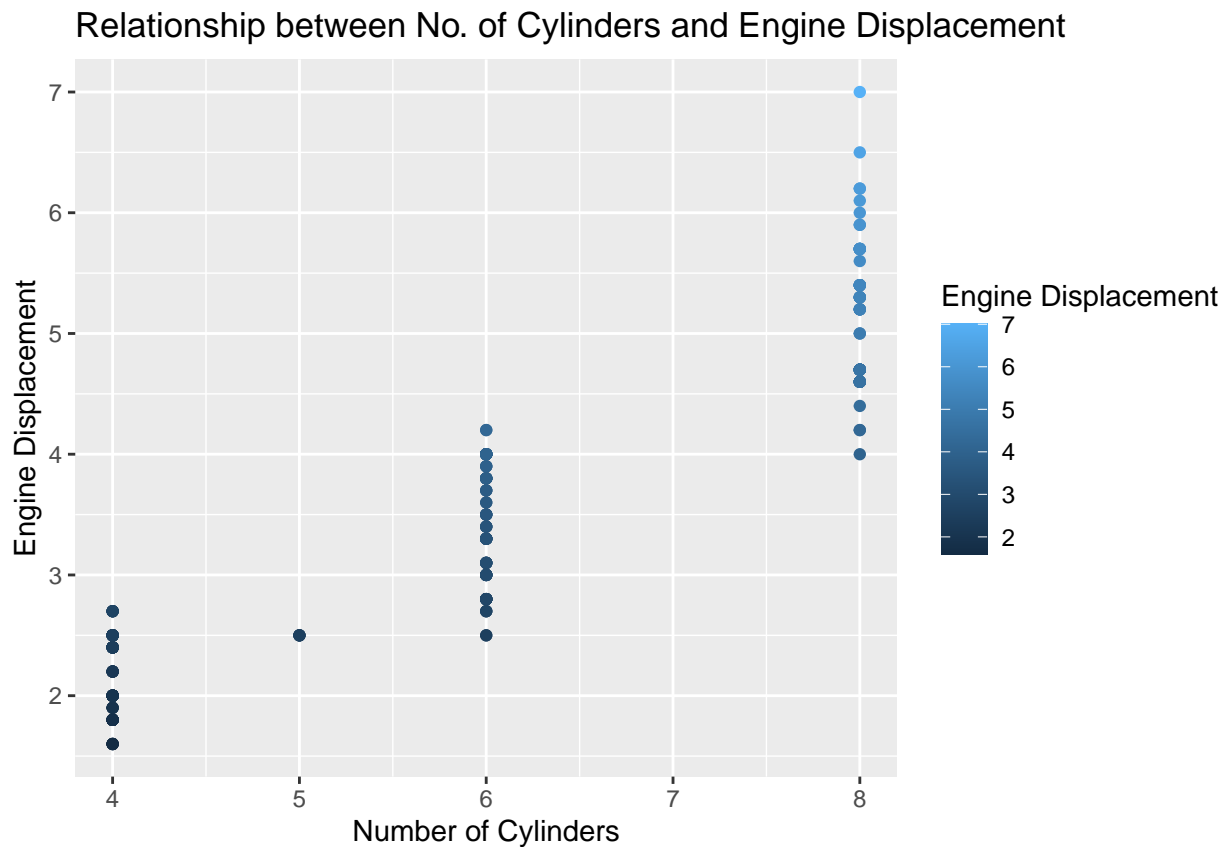
```
library(ggplot2)
top_20 <- head(mpg, 20)
ggplot(top_20, aes(x = model, fill = factor(model))) +
  geom_bar() +
  labs(title = "Number of Cars by Model (Top 20 Observations)",
       x = "Number of Cars",
       y = "Car Model") +
  scale_fill_discrete(name = "Car Model") +
  coord_flip()
```



#5

*# Plot the relationship between cyl - number of cylinders and displ - engine displacement using geom\_point*

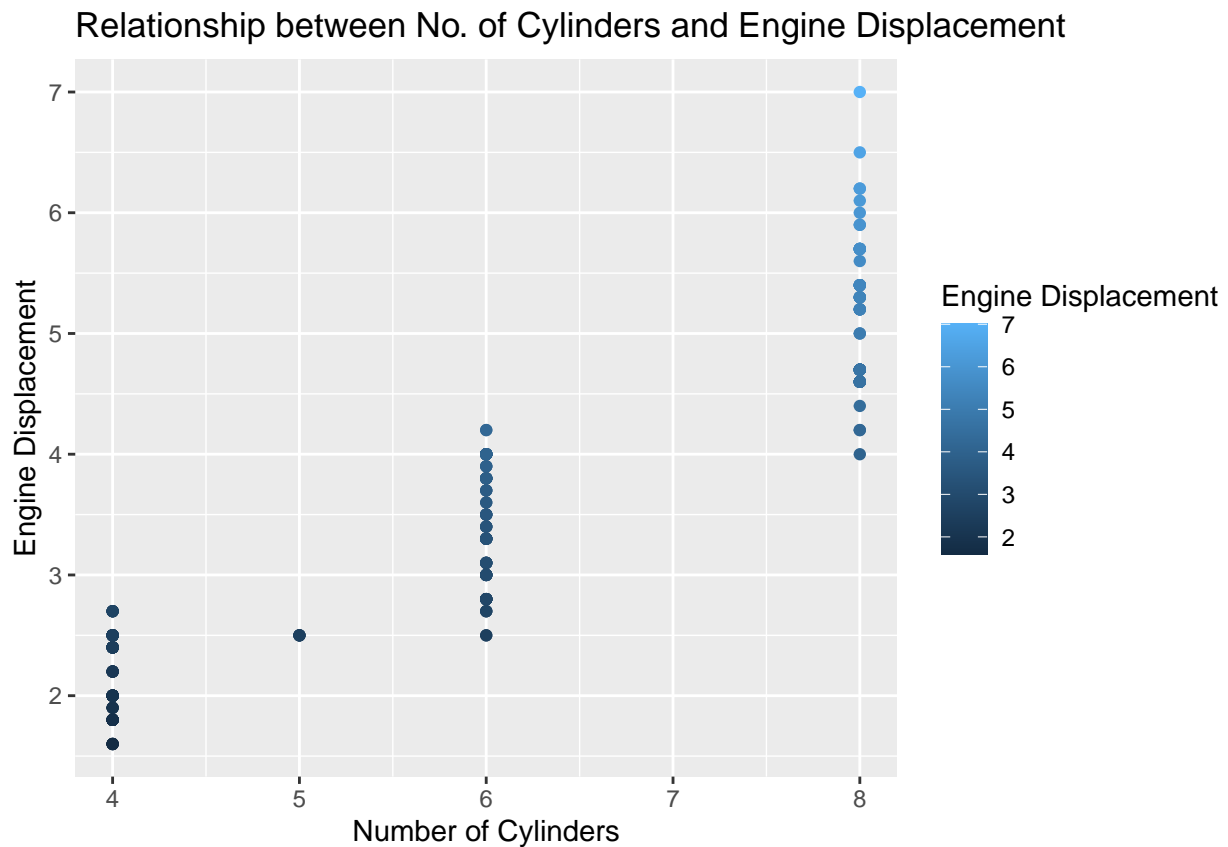
```
library(ggplot2)
ggplot(mpg, aes(x = cyl, y = displ, color = displ)) +
  geom_point() +
  labs(title = "Relationship between No. of Cylinders and Engine Displacement",
       x = "Number of Cylinders",
       y = "Engine Displacement") +
  scale_color_continuous(name = "Engine Displacement")
```



#5a

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

```
library(ggplot2)
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_continuous(name = "Engine Displacement")
```



#6

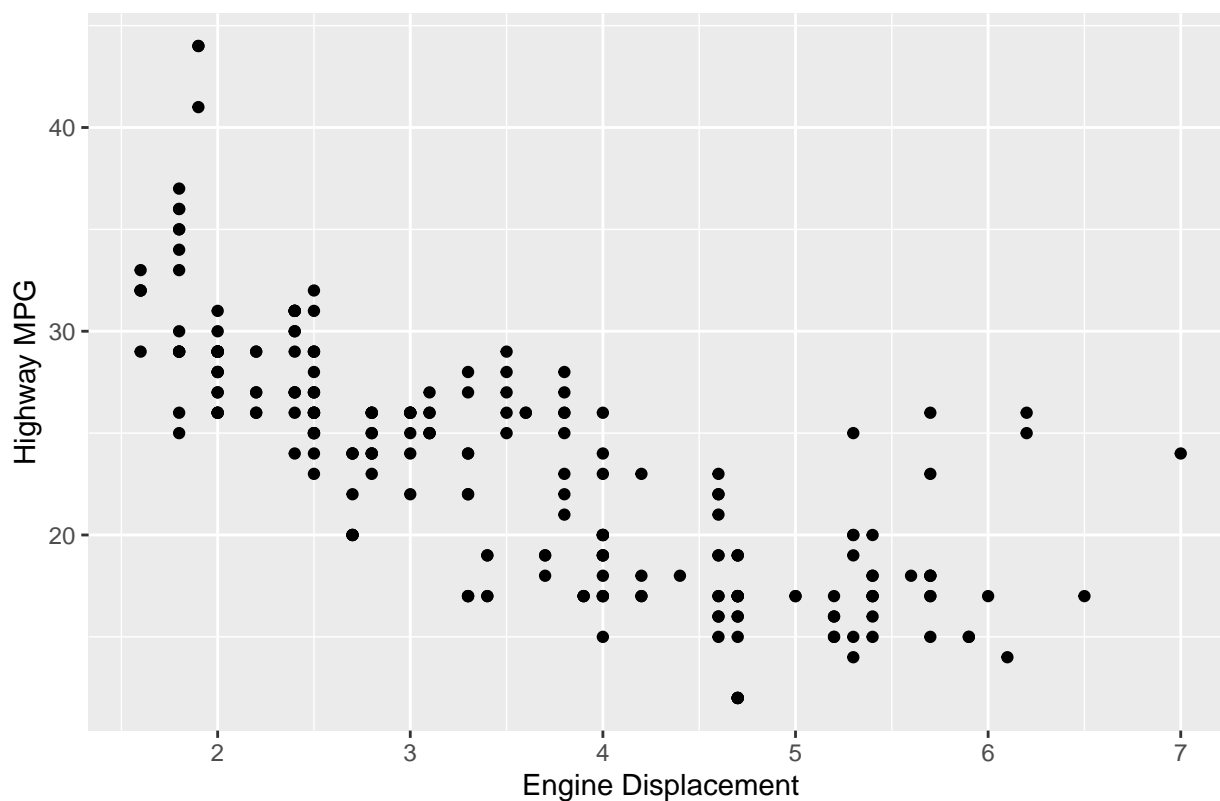
*# Plot the relationship between displ (engine displacement) and hwy(highway miles per gallon). Mapped i*

```
library(ggplot2)
continuous_variable <- mpg$your_continuous_variable

## Warning: Unknown or uninitialised column: `your_continuous_variable`.

ggplot(mpg, aes(x = displ, y = hwy, color = continuous_variable)) +
  geom_point() +
  labs(title = "Relationship between Engine Displacement and Highway MPG",
       x = "Engine Displacement",
       y = "Highway MPG") +
  scale_color_continuous(name = "Your Continuous Variable")
```

Relationship between Engine Displacement and Highway MPG



#6.1

```
library(readr)
traffic <- read_csv("traffic.csv", show_col_types = FALSE)
traffic
```

```
## # A tibble: 48,120 x 4
##   DateTime      Junction Vehicles      ID
##   <dtm>         <dbl>    <dbl>    <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 48,110 more rows
```

#6a

```
observations <- nrow(traffic)
observations
```

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

```
columns <- ncol(traffic)
columns
```

```
## [1] 4
```

```
#6b
```

```
junction1 <- subset(traffic, Junction ==1)
junction1
```

```
## # A tibble: 14,592 x 4
```

```
##   DateTime      Junction Vehicles      ID
##   <dtm>         <dbl>    <dbl>    <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
```

```
junction2 <- subset(traffic, Junction ==2)
junction2
```

```
## # A tibble: 14,592 x 4
```

```
##   DateTime      Junction Vehicles      ID
##   <dtm>         <dbl>    <dbl>    <dbl>
## 1 2015-11-01 00:00:00      2       6 20151101002
## 2 2015-11-01 01:00:00      2       6 20151101012
## 3 2015-11-01 02:00:00      2       5 20151101022
## 4 2015-11-01 03:00:00      2       6 20151101032
## 5 2015-11-01 04:00:00      2       7 20151101042
## 6 2015-11-01 05:00:00      2       2 20151101052
## 7 2015-11-01 06:00:00      2       4 20151101062
## 8 2015-11-01 07:00:00      2       4 20151101072
## 9 2015-11-01 08:00:00      2       3 20151101082
##10 2015-11-01 09:00:00      2       3 20151101092
## # i 14,582 more rows
```

```
junction3 <- subset(traffic, Junction ==3)
junction3
```

```
## # A tibble: 14,592 x 4
```

```
##   DateTime      Junction Vehicles      ID
##   <dtm>         <dbl>    <dbl>    <dbl>
## 1 2015-11-01 00:00:00      3       9 20151101003
## 2 2015-11-01 01:00:00      3       7 20151101013
## 3 2015-11-01 02:00:00      3       5 20151101023
## 4 2015-11-01 03:00:00      3       1 20151101033
## 5 2015-11-01 04:00:00      3       2 20151101043
## 6 2015-11-01 05:00:00      3       2 20151101053
## 7 2015-11-01 06:00:00      3       3 20151101063
## 8 2015-11-01 07:00:00      3       4 20151101073
## 9 2015-11-01 08:00:00      3       3 20151101083
##10 2015-11-01 09:00:00      3       6 20151101093
## # i 14,582 more rows
```

```
junction4 <- subset(traffic, Junction ==4)
junction4
```

```
## # A tibble: 4,344 x 4
##   DateTime      Junction Vehicles      ID
##   <dtm>         <dbl>    <dbl>    <dbl>
## 1 2017-01-01 00:00:00      4      3 20170101004
## 2 2017-01-01 01:00:00      4      1 20170101014
## 3 2017-01-01 02:00:00      4      4 20170101024
## 4 2017-01-01 03:00:00      4      4 20170101034
## 5 2017-01-01 04:00:00      4      2 20170101044
## 6 2017-01-01 05:00:00      4      1 20170101054
## 7 2017-01-01 06:00:00      4      1 20170101064
## 8 2017-01-01 07:00:00      4      4 20170101074
## 9 2017-01-01 08:00:00      4      4 20170101084
## 10 2017-01-01 09:00:00      4      2 20170101094
## # i 4,334 more rows
```

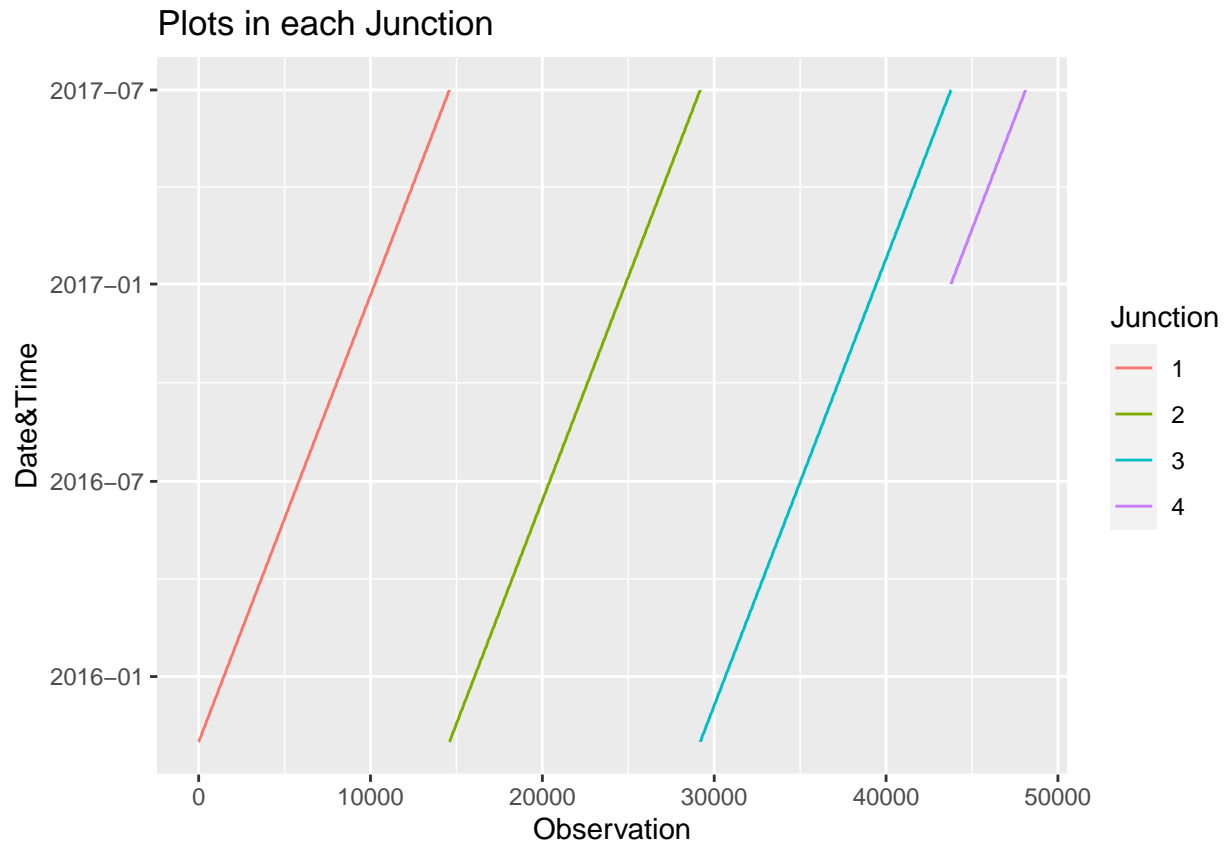
```
#6c
```

```
# Assuming 'traffic' is your dataset
```

```
library(ggplot2)
```

```
# Plot each junction using geom_line()
```

```
ggplot(traffic, aes(x = seq_along(Junction), y = DateTime, group = Junction, color = factor(Junction)))
  geom_line() +
  labs(title = "Plots in each Junction",
       x = "Observation",
       y = "Date&Time") +
  scale_color_discrete(name = "Junction")
```



#7

```
#install.packages("readxl")
library(readxl)
alexafile <- read_excel("alexa_file.xlsx")
alexafile
```

## # A tibble: 3,150 x 5

	rating	date	variation	verified_reviews	feedback
	<dbl>	<dtm>	<chr>	<chr>	<dbl>
## 1	5	2018-07-31 00:00:00	Charcoal Fabric	Love my Echo!	1
## 2	5	2018-07-31 00:00:00	Charcoal Fabric	Loved it!	1
## 3	4	2018-07-31 00:00:00	Walnut Finish	Sometimes while play~	1
## 4	5	2018-07-31 00:00:00	Charcoal Fabric	I have had a lot of ~	1
## 5	5	2018-07-31 00:00:00	Charcoal Fabric	Music	1
## 6	5	2018-07-31 00:00:00	Heather Gray Fabric	I received the echo ~	1
## 7	3	2018-07-31 00:00:00	Sandstone Fabric	Without having a cel~	1
## 8	5	2018-07-31 00:00:00	Charcoal Fabric	I think this is the ~	1
## 9	5	2018-07-30 00:00:00	Heather Gray Fabric	looks great	1
## 10	5	2018-07-30 00:00:00	Heather Gray Fabric	Love it! I've listen~	1

## # i 3,140 more rows

#7a

```
num_rows <- nrow(alexafile)
num_columns <- ncol(alexafile)
num_rows
```

## [1] 3150



```
num_columns
```

```
## [1] 5
```

```
#7b
```

```
library(dplyr)
```

```
output <- alexafile %>%  
  group_by(variation) %>%  
  summarize(total_count = n())
```

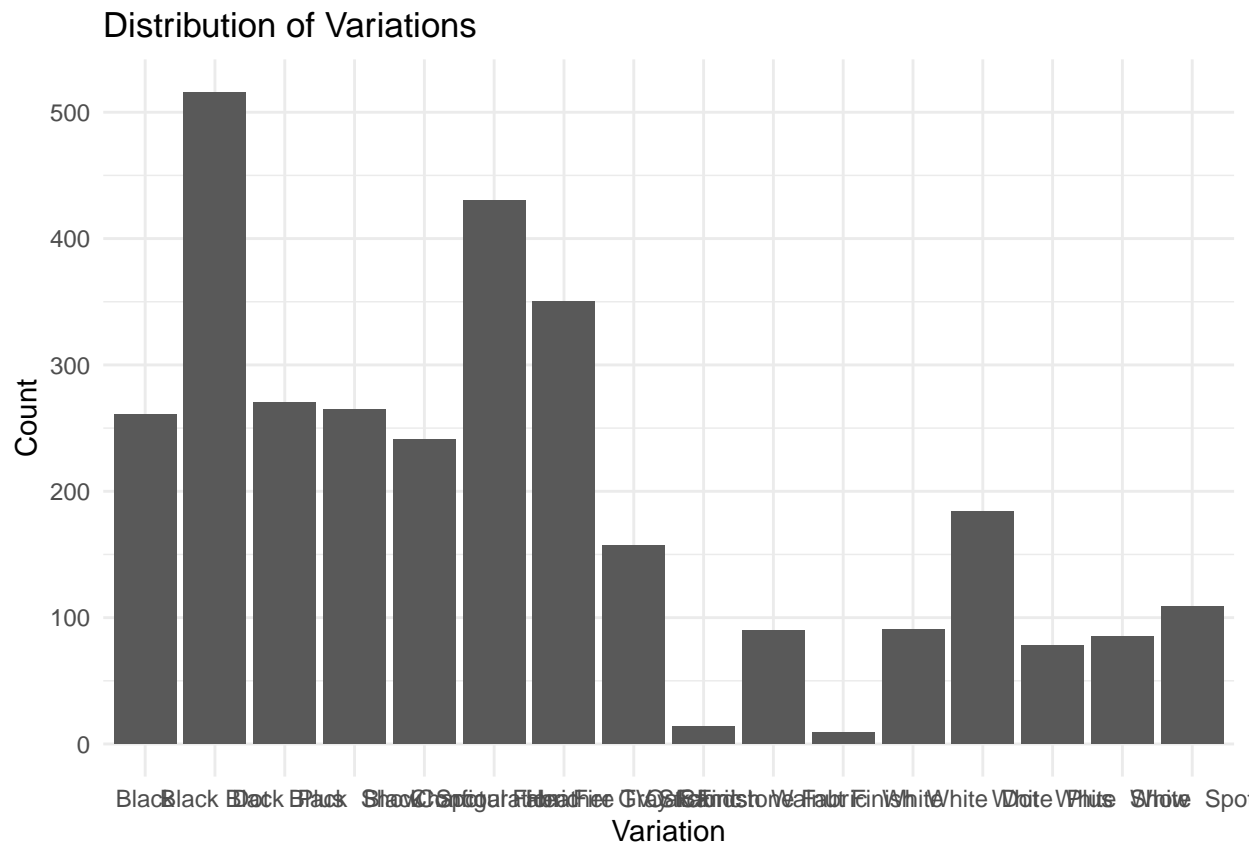
```
print(output)
```

```
## # A tibble: 16 x 2  
##   variation                total_count  
##   <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
```

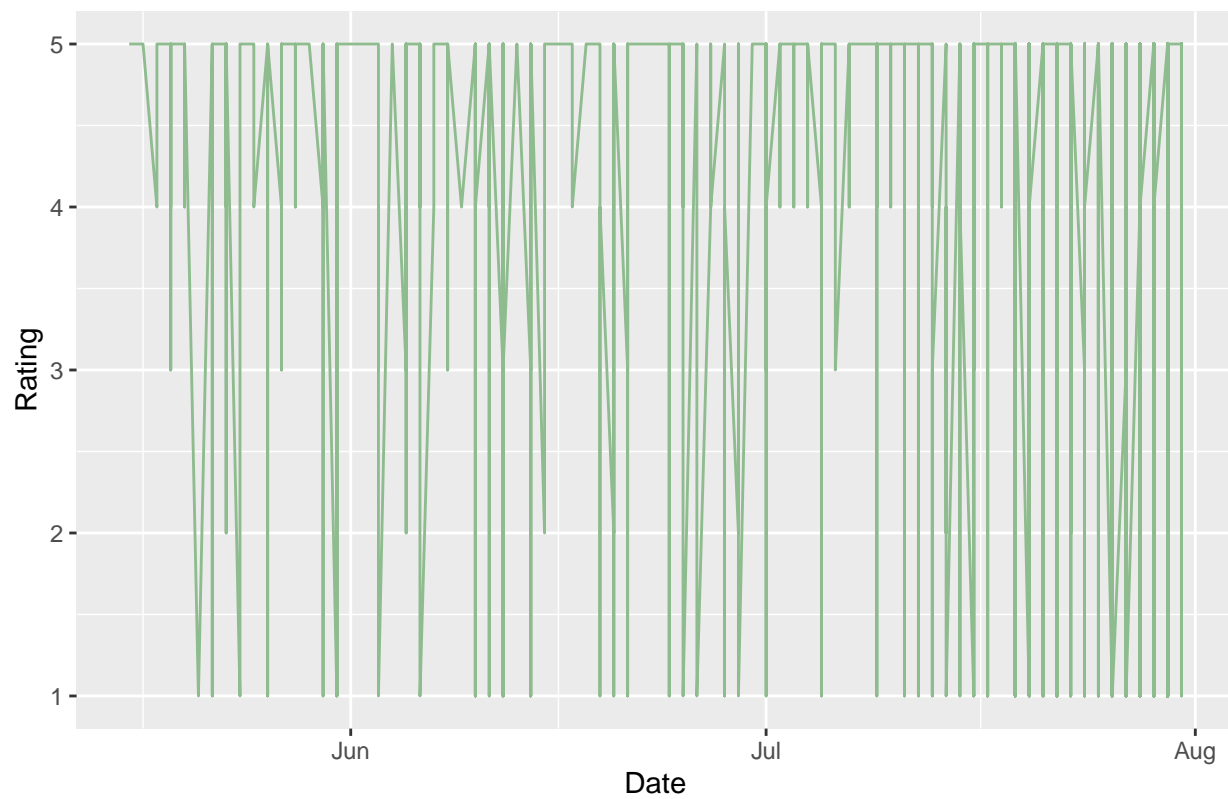
```
library(ggplot2)
```

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



```
#7d
ggplot(alexafile, aes(x = date, y= rating)) +
  geom_line(color= "darkseagreen") +
  labs (title = "Date and the number of verified reviews",
        x = "Date",
        y = "Rating")
```

Date and the number of verified reviews



#7e

```
ggplot(alexafile, aes(x = variation, y = rating)) +
  geom_boxplot(fill = "coral", color = "slategray") +
  labs(title = "Relationship Between Variations and Ratings",
        x = "Variations",
        y = "Ratings") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

# Relationship Between Variations and Ratings

