

RWorksheet_Olivo#4c

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1. Use the dataset mpg Download and open the mpg file. Upload it to your OWN environment

1a. Show your solutions on how to import a csv file into the environment.

```
library(readr)
```

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

1b. Which variables from mpg dataset are categorical?

```
str(mpg_file)
```

```
## '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" ...
```

```
# manufacturer, model, trans, drv, fl, class variables are categorical
```

1c. Which are continuous variables?

```
str(mpg_file)
```

```
## '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" ...
```

```
# X, displ, year, cyl, cty, hwy are continuous variables
```

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

```
manufacturer_asTable <- table(mpg_file$manufacturer)
manufacturer_most_models <- names(manufacturer_asTable)[which.max(manufacturer_asTable)]

manufacturer_most_models
```

```
## [1] "dodge"
```

```
# dodge manufacturer has the most models
```

```
model_asTable <- table(mpg_file$model)
model_most_vars <- names(model_asTable)[which.max(model_asTable)]

model_most_vars
```

```
## [1] "caravan 2wd"
```

```
# caravan 2wd has the most variations
```

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

```
#install.packages("dplyr")
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
```

```
manufacturers_models <- data.frame(Manufacturer = mpg_file$manufacturer, Model = mpg_file$model)
manufacturers_models
```

```
##      Manufacturer      Model
## 1         audi         a4
## 2         audi         a4
## 3         audi         a4
## 4         audi         a4
## 5         audi         a4
## 6         audi         a4
## 7         audi         a4
## 8         audi    a4 quattro
## 9         audi    a4 quattro
## 10        audi    a4 quattro
## 11        audi    a4 quattro
## 12        audi    a4 quattro
## 13        audi    a4 quattro
## 14        audi    a4 quattro
## 15        audi    a4 quattro
```

## 16	audi	a6 quattro
## 17	audi	a6 quattro
## 18	audi	a6 quattro
## 19	chevrolet	c1500 suburban 2wd
## 20	chevrolet	c1500 suburban 2wd
## 21	chevrolet	c1500 suburban 2wd
## 22	chevrolet	c1500 suburban 2wd
## 23	chevrolet	c1500 suburban 2wd
## 24	chevrolet	corvette
## 25	chevrolet	corvette
## 26	chevrolet	corvette
## 27	chevrolet	corvette
## 28	chevrolet	corvette
## 29	chevrolet	k1500 tahoe 4wd
## 30	chevrolet	k1500 tahoe 4wd
## 31	chevrolet	k1500 tahoe 4wd
## 32	chevrolet	k1500 tahoe 4wd
## 33	chevrolet	malibu
## 34	chevrolet	malibu
## 35	chevrolet	malibu
## 36	chevrolet	malibu
## 37	chevrolet	malibu
## 38	dodge	caravan 2wd
## 39	dodge	caravan 2wd
## 40	dodge	caravan 2wd
## 41	dodge	caravan 2wd
## 42	dodge	caravan 2wd
## 43	dodge	caravan 2wd
## 44	dodge	caravan 2wd
## 45	dodge	caravan 2wd
## 46	dodge	caravan 2wd
## 47	dodge	caravan 2wd
## 48	dodge	caravan 2wd
## 49	dodge	dakota pickup 4wd
## 50	dodge	dakota pickup 4wd
## 51	dodge	dakota pickup 4wd
## 52	dodge	dakota pickup 4wd
## 53	dodge	dakota pickup 4wd
## 54	dodge	dakota pickup 4wd
## 55	dodge	dakota pickup 4wd
## 56	dodge	dakota pickup 4wd
## 57	dodge	dakota pickup 4wd
## 58	dodge	durango 4wd
## 59	dodge	durango 4wd
## 60	dodge	durango 4wd
## 61	dodge	durango 4wd
## 62	dodge	durango 4wd
## 63	dodge	durango 4wd
## 64	dodge	durango 4wd
## 65	dodge	ram 1500 pickup 4wd
## 66	dodge	ram 1500 pickup 4wd
## 67	dodge	ram 1500 pickup 4wd
## 68	dodge	ram 1500 pickup 4wd
## 69	dodge	ram 1500 pickup 4wd

## 70	dodge	ram 1500 pickup 4wd
## 71	dodge	ram 1500 pickup 4wd
## 72	dodge	ram 1500 pickup 4wd
## 73	dodge	ram 1500 pickup 4wd
## 74	dodge	ram 1500 pickup 4wd
## 75	ford	expedition 2wd
## 76	ford	expedition 2wd
## 77	ford	expedition 2wd
## 78	ford	explorer 4wd
## 79	ford	explorer 4wd
## 80	ford	explorer 4wd
## 81	ford	explorer 4wd
## 82	ford	explorer 4wd
## 83	ford	explorer 4wd
## 84	ford	f150 pickup 4wd
## 85	ford	f150 pickup 4wd
## 86	ford	f150 pickup 4wd
## 87	ford	f150 pickup 4wd
## 88	ford	f150 pickup 4wd
## 89	ford	f150 pickup 4wd
## 90	ford	f150 pickup 4wd
## 91	ford	mustang
## 92	ford	mustang
## 93	ford	mustang
## 94	ford	mustang
## 95	ford	mustang
## 96	ford	mustang
## 97	ford	mustang
## 98	ford	mustang
## 99	ford	mustang
## 100	honda	civic
## 101	honda	civic
## 102	honda	civic
## 103	honda	civic
## 104	honda	civic
## 105	honda	civic
## 106	honda	civic
## 107	honda	civic
## 108	honda	civic
## 109	hyundai	sonata
## 110	hyundai	sonata
## 111	hyundai	sonata
## 112	hyundai	sonata
## 113	hyundai	sonata
## 114	hyundai	sonata
## 115	hyundai	sonata
## 116	hyundai	tiburon
## 117	hyundai	tiburon
## 118	hyundai	tiburon
## 119	hyundai	tiburon
## 120	hyundai	tiburon
## 121	hyundai	tiburon
## 122	hyundai	tiburon
## 123	jeep	grand cherokee 4wd

## 124	jeep	grand cherokee 4wd
## 125	jeep	grand cherokee 4wd
## 126	jeep	grand cherokee 4wd
## 127	jeep	grand cherokee 4wd
## 128	jeep	grand cherokee 4wd
## 129	jeep	grand cherokee 4wd
## 130	jeep	grand cherokee 4wd
## 131	land rover	range rover
## 132	land rover	range rover
## 133	land rover	range rover
## 134	land rover	range rover
## 135	lincoln	navigator 2wd
## 136	lincoln	navigator 2wd
## 137	lincoln	navigator 2wd
## 138	mercury	mountaineer 4wd
## 139	mercury	mountaineer 4wd
## 140	mercury	mountaineer 4wd
## 141	mercury	mountaineer 4wd
## 142	nissan	altima
## 143	nissan	altima
## 144	nissan	altima
## 145	nissan	altima
## 146	nissan	altima
## 147	nissan	altima
## 148	nissan	maxima
## 149	nissan	maxima
## 150	nissan	maxima
## 151	nissan	pathfinder 4wd
## 152	nissan	pathfinder 4wd
## 153	nissan	pathfinder 4wd
## 154	nissan	pathfinder 4wd
## 155	pontiac	grand prix
## 156	pontiac	grand prix
## 157	pontiac	grand prix
## 158	pontiac	grand prix
## 159	pontiac	grand prix
## 160	subaru	forester awd
## 161	subaru	forester awd
## 162	subaru	forester awd
## 163	subaru	forester awd
## 164	subaru	forester awd
## 165	subaru	forester awd
## 166	subaru	impreza awd
## 167	subaru	impreza awd
## 168	subaru	impreza awd
## 169	subaru	impreza awd
## 170	subaru	impreza awd
## 171	subaru	impreza awd
## 172	subaru	impreza awd
## 173	subaru	impreza awd
## 174	toyota	4runner 4wd
## 175	toyota	4runner 4wd
## 176	toyota	4runner 4wd
## 177	toyota	4runner 4wd

## 178	toyota	4runner 4wd
## 179	toyota	4runner 4wd
## 180	toyota	camry
## 181	toyota	camry
## 182	toyota	camry
## 183	toyota	camry
## 184	toyota	camry
## 185	toyota	camry
## 186	toyota	camry
## 187	toyota	camry solara
## 188	toyota	camry solara
## 189	toyota	camry solara
## 190	toyota	camry solara
## 191	toyota	camry solara
## 192	toyota	camry solara
## 193	toyota	camry solara
## 194	toyota	corolla
## 195	toyota	corolla
## 196	toyota	corolla
## 197	toyota	corolla
## 198	toyota	corolla
## 199	toyota	land cruiser wagon 4wd
## 200	toyota	land cruiser wagon 4wd
## 201	toyota	toyota tacoma 4wd
## 202	toyota	toyota tacoma 4wd
## 203	toyota	toyota tacoma 4wd
## 204	toyota	toyota tacoma 4wd
## 205	toyota	toyota tacoma 4wd
## 206	toyota	toyota tacoma 4wd
## 207	toyota	toyota tacoma 4wd
## 208	volkswagen	gti
## 209	volkswagen	gti
## 210	volkswagen	gti
## 211	volkswagen	gti
## 212	volkswagen	gti
## 213	volkswagen	jetta
## 214	volkswagen	jetta
## 215	volkswagen	jetta
## 216	volkswagen	jetta
## 217	volkswagen	jetta
## 218	volkswagen	jetta
## 219	volkswagen	jetta
## 220	volkswagen	jetta
## 221	volkswagen	jetta
## 222	volkswagen	new beetle
## 223	volkswagen	new beetle
## 224	volkswagen	new beetle
## 225	volkswagen	new beetle
## 226	volkswagen	new beetle
## 227	volkswagen	new beetle
## 228	volkswagen	passat
## 229	volkswagen	passat
## 230	volkswagen	passat
## 231	volkswagen	passat

```
## 232 volkswagen      passat
## 233 volkswagen      passat
## 234 volkswagen      passat
```

```
unique_mods <- unique(manufacturers_models)
unique_mods
```

```
##      Manufacturer      Model
## 1         audi         a4
## 8         audi      a4 quattro
## 16        audi      a6 quattro
## 19   chevrolet  c1500 suburban 2wd
## 24   chevrolet      corvette
## 29   chevrolet  k1500 tahoe 4wd
## 33   chevrolet      malibu
## 38        dodge      caravan 2wd
## 49        dodge  dakota pickup 4wd
## 58        dodge      durango 4wd
## 65        dodge  ram 1500 pickup 4wd
## 75         ford      expedition 2wd
## 78         ford      explorer 4wd
## 84         ford      f150 pickup 4wd
## 91         ford      mustang
## 100        honda      civic
## 109       hyundai      sonata
## 116       hyundai      tiburon
## 123        jeep  grand cherokee 4wd
## 131  land rover      range rover
## 135       lincoln      navigator 2wd
## 138       mercury      mountaineer 4wd
## 142        nissan      altima
## 148        nissan      maxima
## 151        nissan      pathfinder 4wd
## 155       pontiac      grand prix
## 160       subaru      forester awd
## 166       subaru      impreza awd
## 174       toyota      4runner 4wd
## 180       toyota      camry
## 187       toyota      camry solara
## 194       toyota      corolla
## 199       toyota  land cruiser wagon 4wd
## 201       toyota      toyota tacoma 4wd
## 208  volkswagen      gti
## 213  volkswagen      jetta
## 222  volkswagen      new beetle
## 228  volkswagen      passat
```

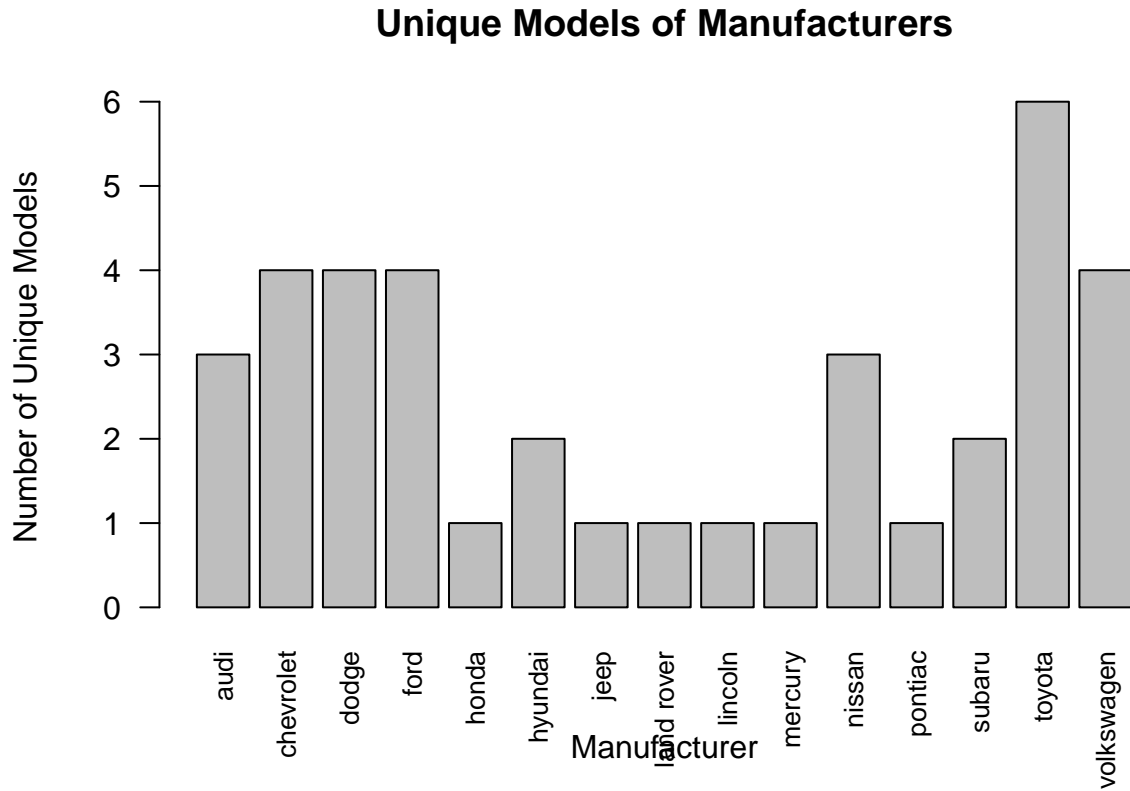
```
unique_mods_factor <- factoredManufacturer <- as.factor(unique_mods$Manufacturer)
```

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

```
#install.packages("ggplot2")
library(ggplot2)

#install.packages("dplyr")
library(dplyr)
```

```
unique_plot <- plot(as.factor(factoredManufacturer),
  main = "Unique Models of Manufacturers",
  xlab = "Manufacturer",
  ylab = "Number of Unique Models",
  cex.names = 0.8, las = 2)
```



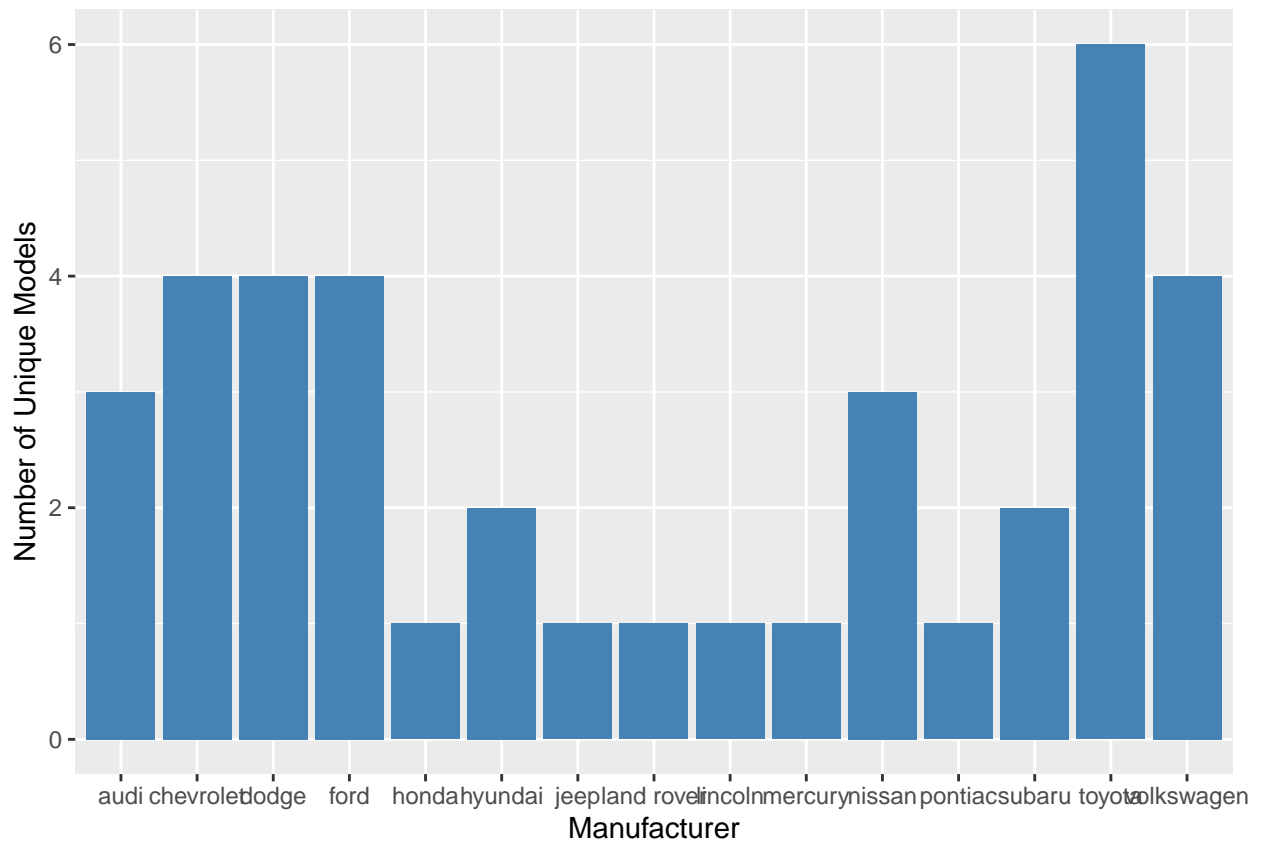
```
unique_count <- unique_mods %>%
  count(unique_mods$Manufacturer)
unique_count
```

```
##   unique_mods$Manufacturer n
## 1             audi 3
## 2        chevrolet 4
## 3             dodge 4
## 4             ford 4
## 5             honda 1
## 6          hyundai 2
## 7             jeep 1
## 8        land rover 1
## 9          lincoln 1
## 10          mercury 1
## 11            nissan 3
## 12          pontiac 1
## 13          subaru 2
## 14            toyota 6
## 15        volkswagen 4
```

```
ggplot(unique_count, aes(x = `unique_mods$Manufacturer`, y = n)) +
  geom_bar(stat = "identity", fill = "steelblue") +
```



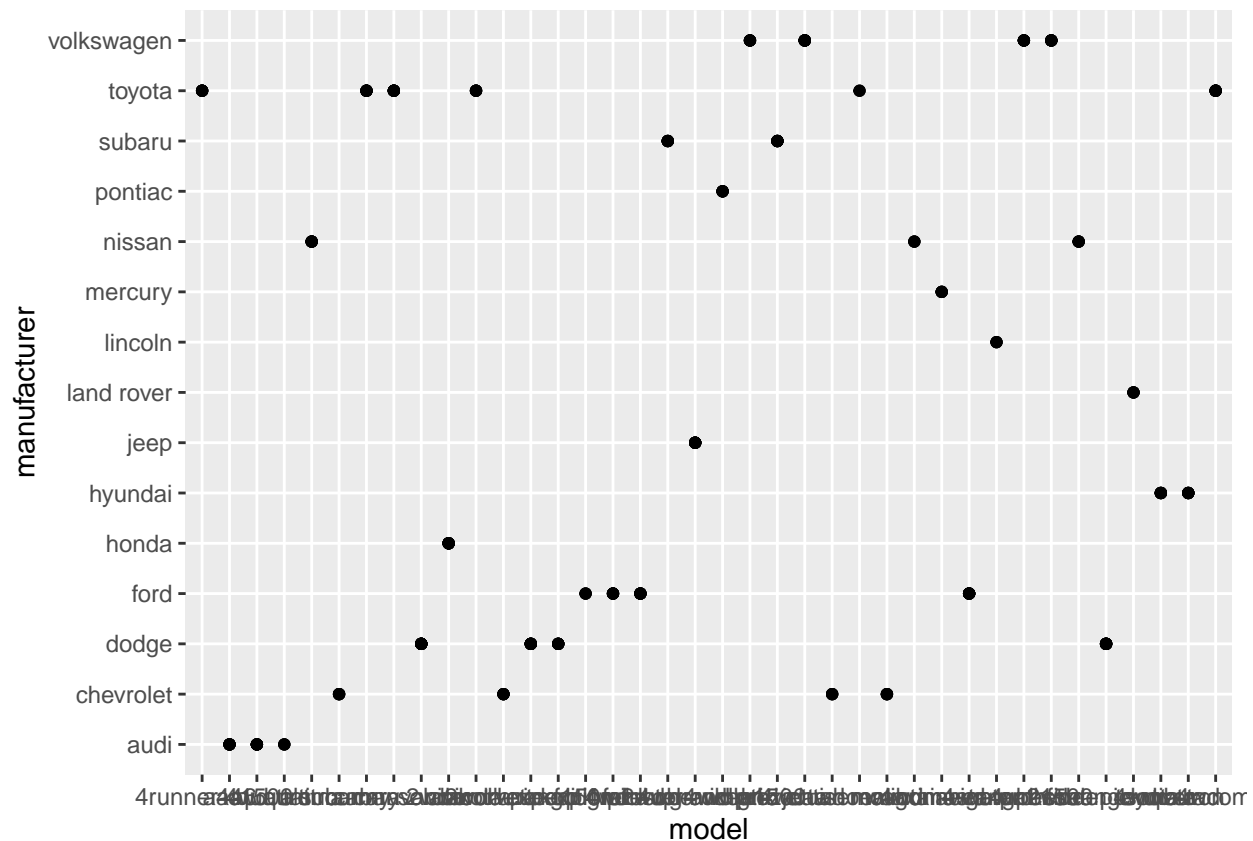
```
labs(x = "Manufacturer", y = "Number of Unique Models")
```



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

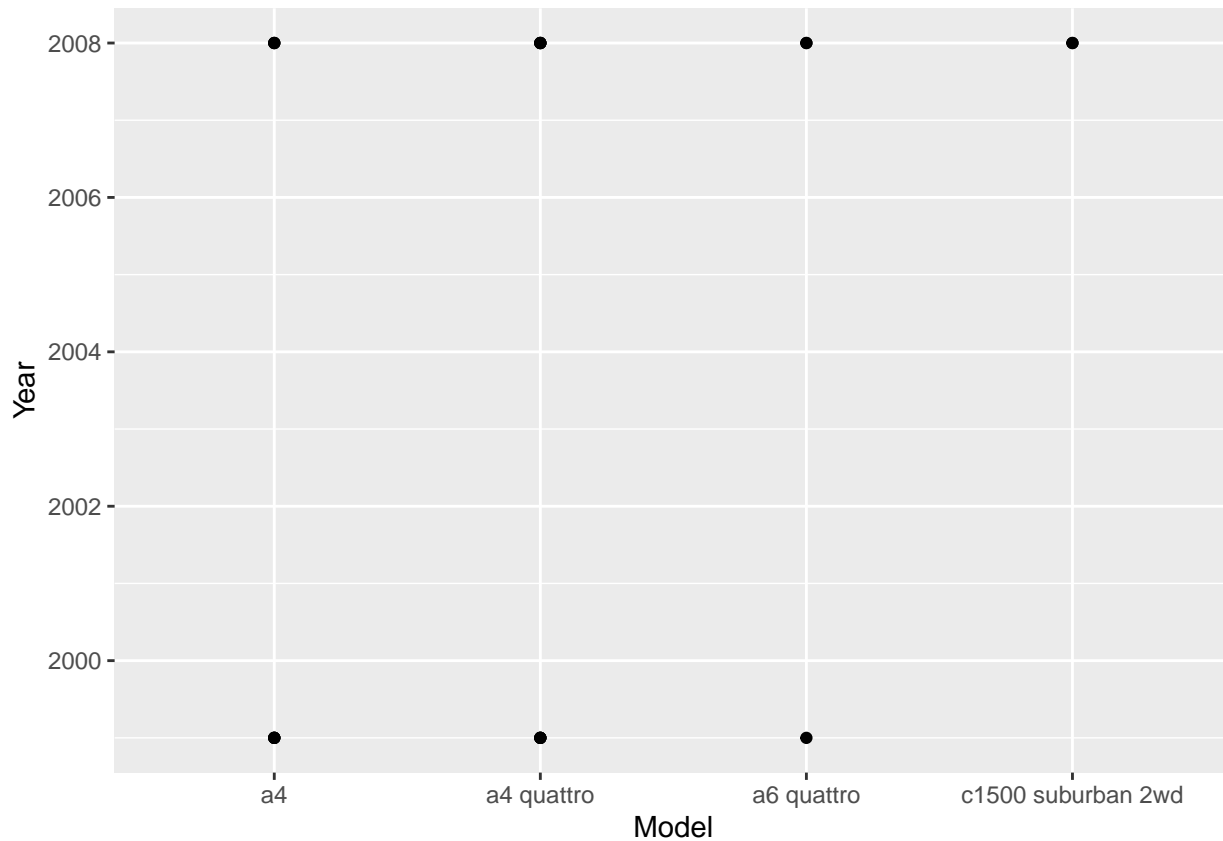
2a. What does `ggplot(mpg, aes(model, manufacturer)) + geom_point()` show?

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



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

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



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

```
library(dplyr)

model_car_count <- mpg_file %>%
  group_by(model) %>%
  summarize(number_of_cars = n())

model_car_count
```

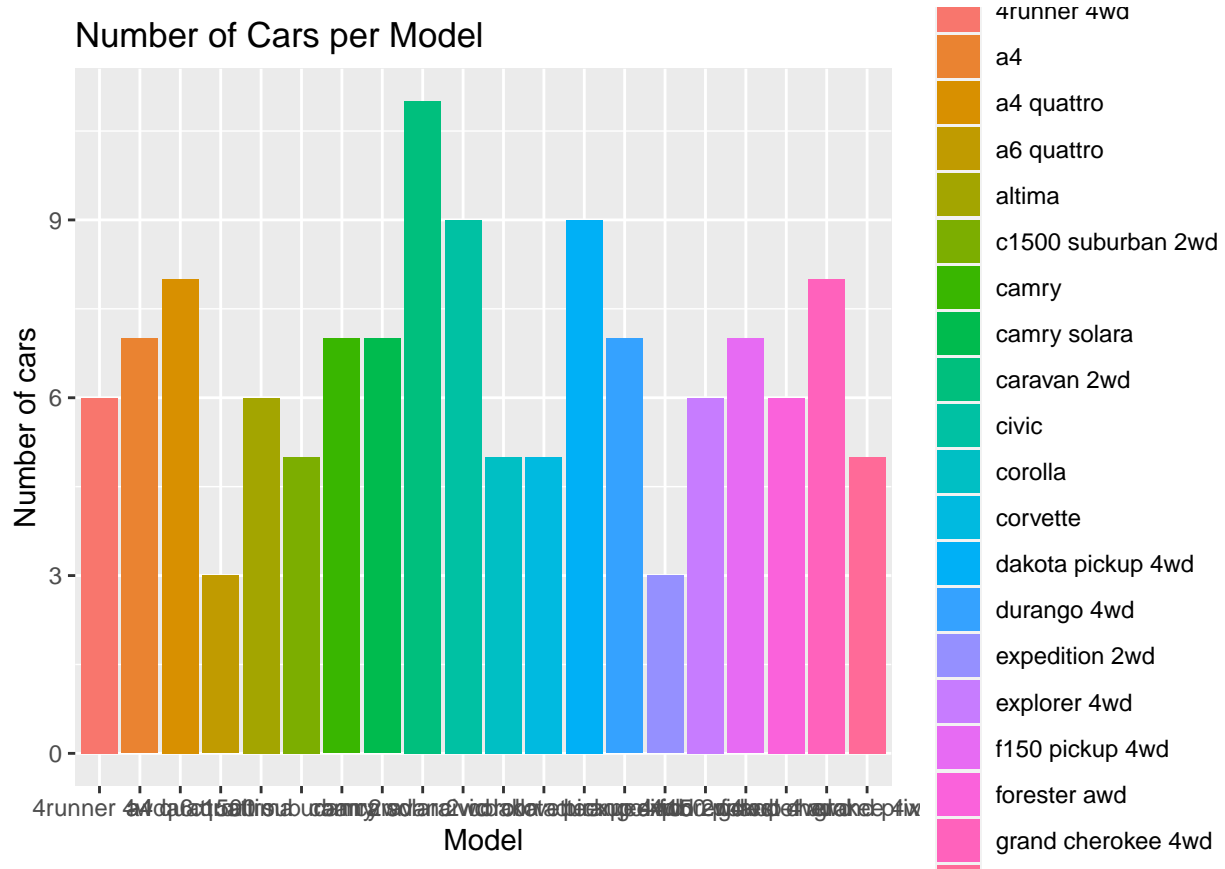
```
## # A tibble: 38 x 2
##   model          number_of_cars
##   <chr>              <int>
## 1 4runner 4wd             6
## 2 a4                     7
## 3 a4 quattro             8
## 4 a6 quattro             3
## 5 altima                 6
## 6 c1500 suburban 2wd     5
## 7 camry                  7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

4a. Plot using `geom_bar()` using the top 20 observations only. The graphs should have a title, labels and colors. Show code and results

```
obs_20 <- head(model_car_count, 20)
```

```
top_20 <- ggplot(obs_20, aes(x = model, y = number_of_cars, fill = model)) + geom_bar(stat = "identity")
```

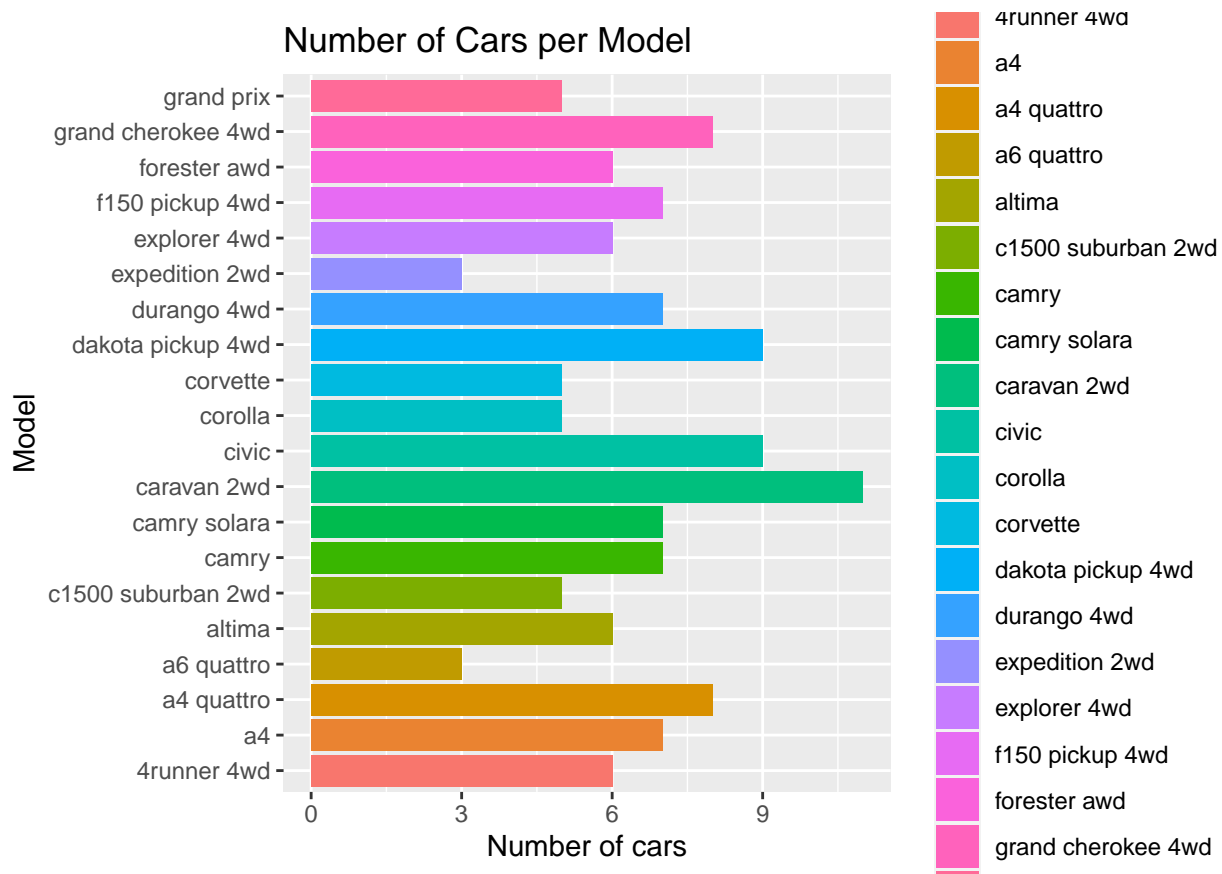
```
top_20
```



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

```
flipped_top_20 <- ggplot(obs_20, aes(x = model, y = number_of_cars, fill = model)) + geom_bar(stat = "identity")
```

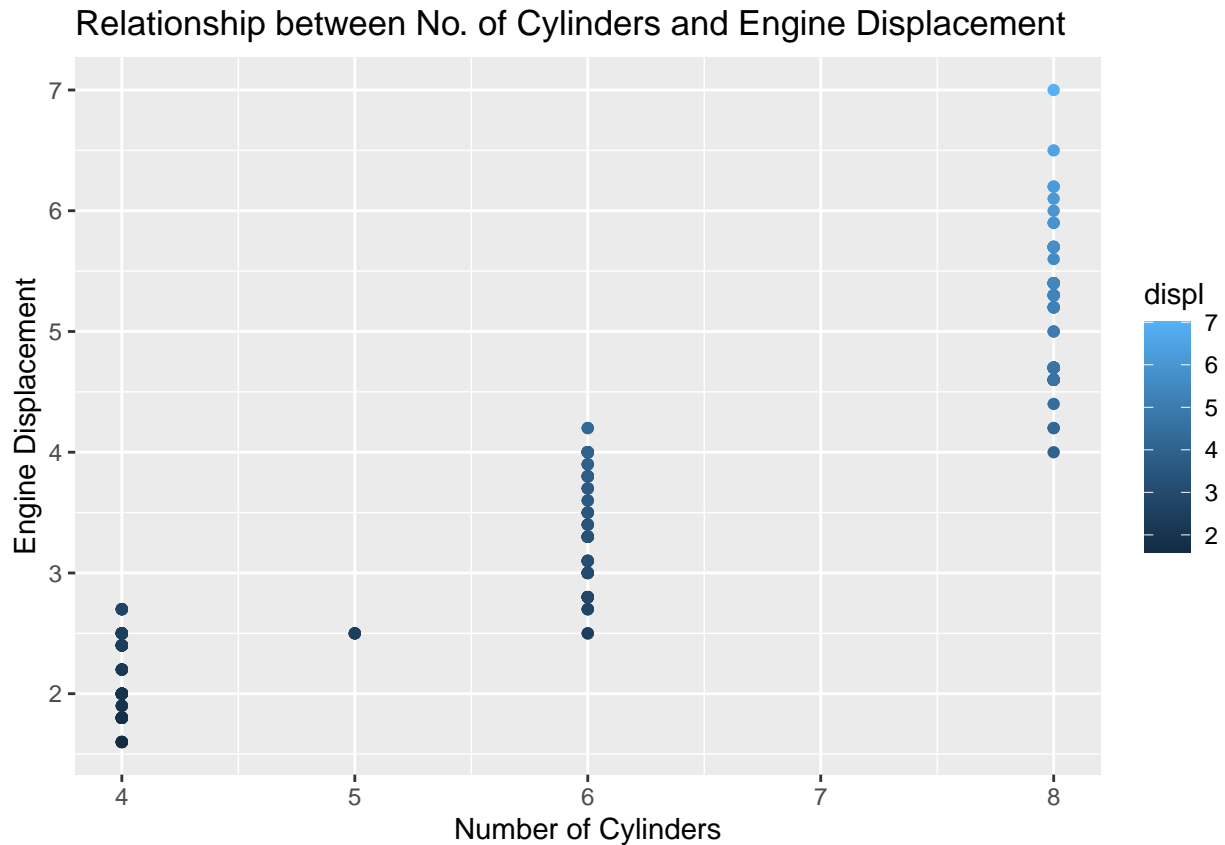
```
flipped_top_20
```



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".

```
cyl_displ_plot <- ggplot(mpg_file, 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")

cyl_displ_plot
```



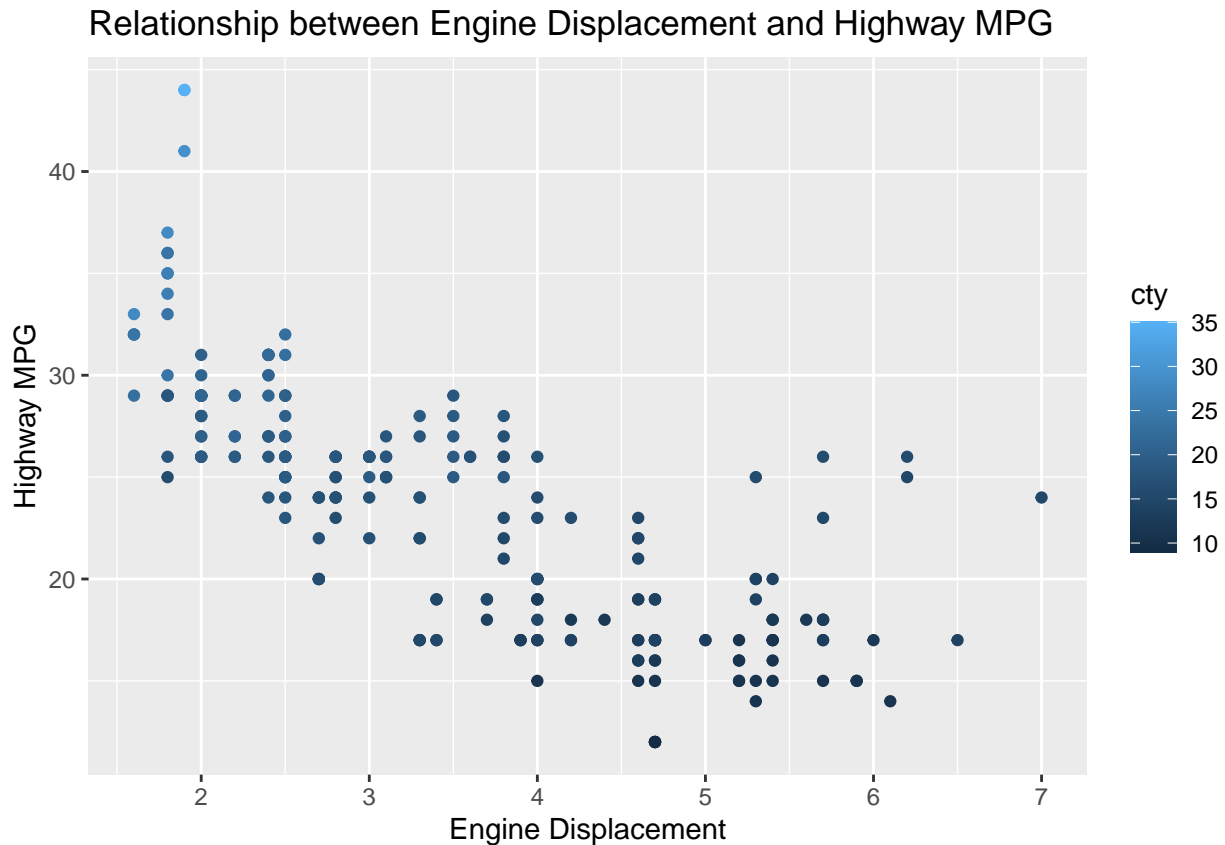
5a. How would you describe its relationship? Show the codes and its result.

```
# It will generate a scatter plot showing the relationship between the number of cylinders and engine d
# As the number of cylinders increases, the engine displacement tends to increase as well. This suggest
```

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. What is its result? Why it produced such output?

```
displ_hwy_plot <- ggplot(mpg_file, aes(x = displ, y = hwy, color = cty)) +
  geom_point() +
  labs(title = "Relationship between Engine Displacement and Highway MPG",
       x = "Engine Displacement",
       y = "Highway MPG")

displ_hwy_plot
```



This is a scatterplot with engine displacement on the x-axis and highway miles per gallon on the y-axis

Using this plot, we can understand the relationship between the displ, hwy, and cty. By mapping the cty

This can provide understanding of the fuel efficiency of vehicle with different engine sizes

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

6a. How many numbers of observation does it have? What are the variables of the traffic dataset the Show your answer.

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

```
num_obs <- nrow(traffic)
num_obs
```

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

```
num_vars <- ncol(traffic)
num_vars
```

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

```
vars <- colnames(traffic)
vars
```

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

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

```

junctions_subset_1 <- subset(traffic, Junction == 1)

junctions_subset_2 <- subset(traffic, Junction == 2)

junctions_subset_3 <- subset(traffic, Junction == 3)

junctions_subset_4 <- subset(traffic, Junction == 4)

```

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

```

junction_1_plot <- ggplot(junctions_subset_1, aes(x = as.Date(junctions_subset_1$DateTime), y = Vehicles)) +
  geom_line() +
  scale_x_date(date_labels = "%b-%Y") + theme(legend.position = "none") +
  labs(title = "Junction 1", x = "Time", y = "Number of Vehicles")

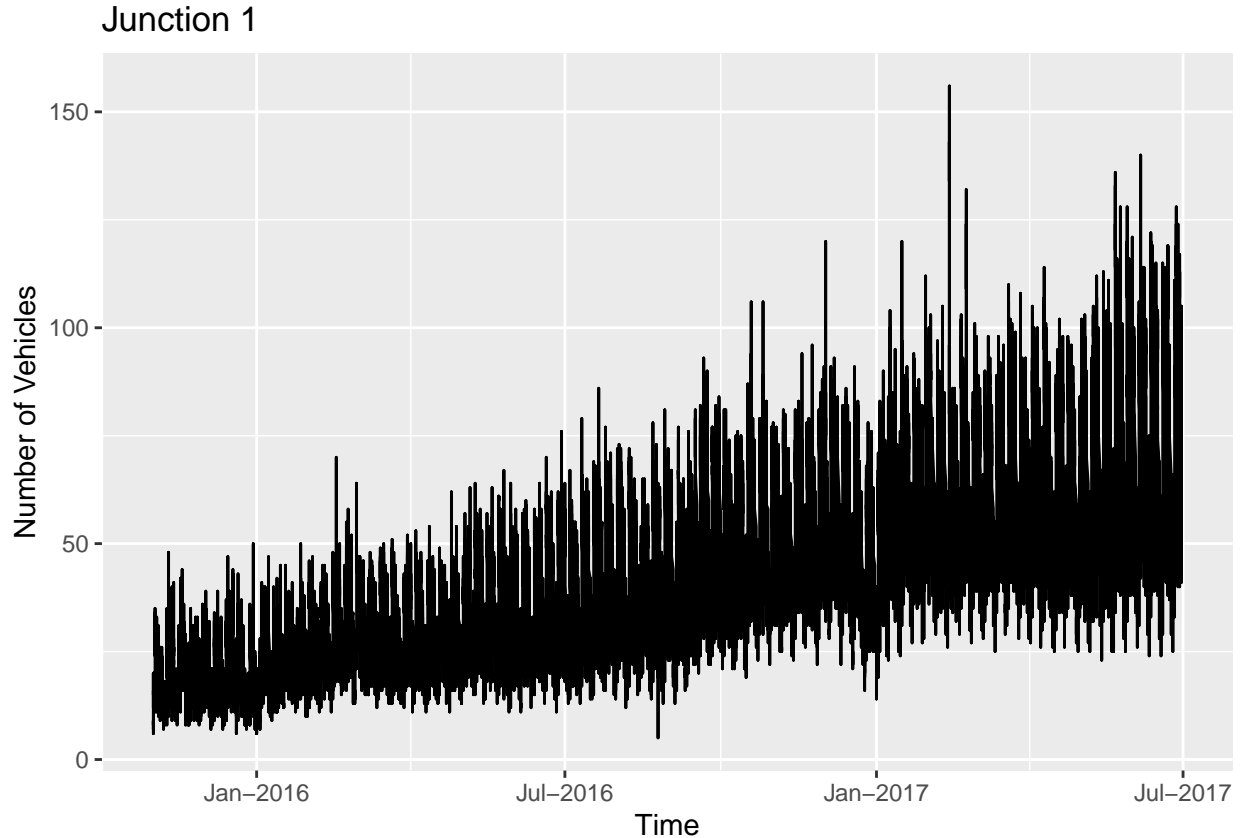
junction_1_plot

```

```

## Warning: Use of `junctions_subset_1$DateTime` is discouraged.
## i Use `DateTime` instead.

```

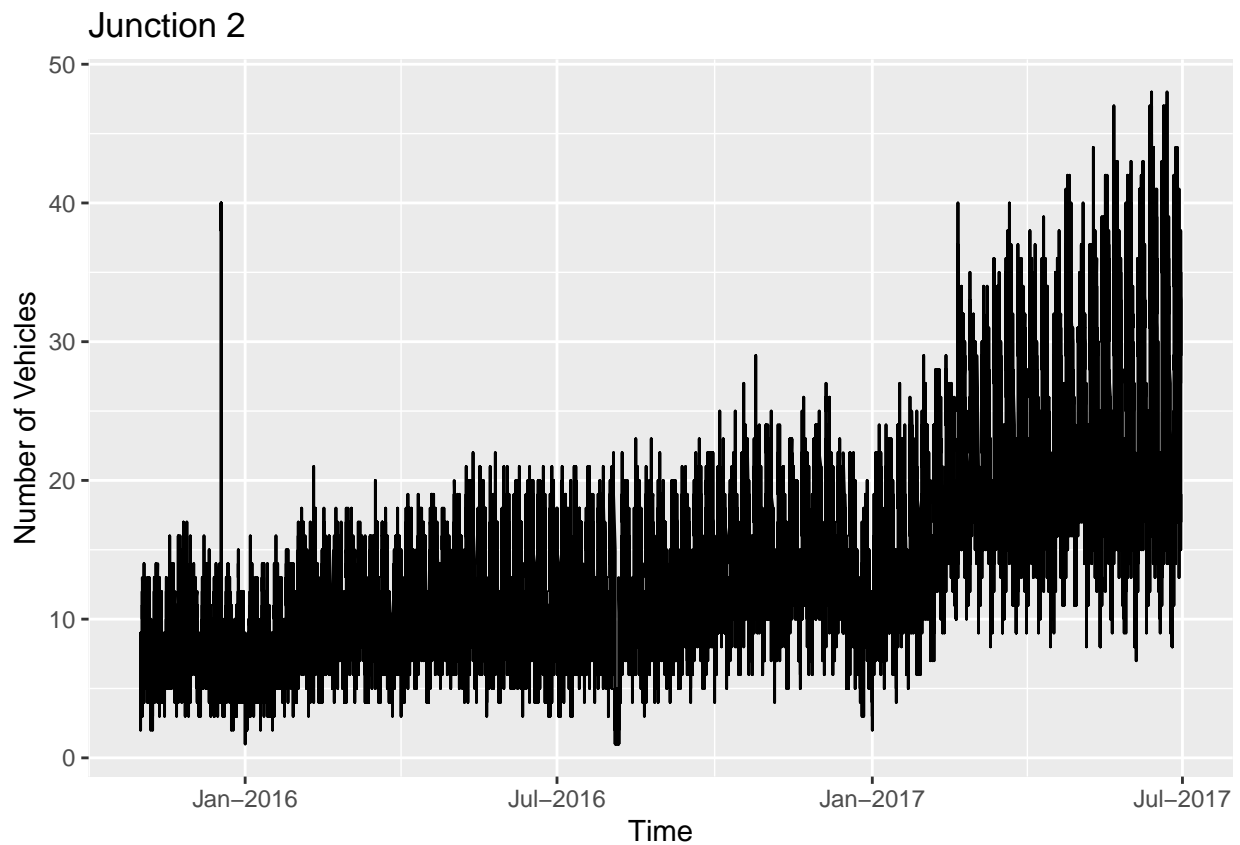


```

junction_2_plot <- ggplot(junctions_subset_2, aes(x = as.Date(junctions_subset_2$DateTime), y = Vehicles)) +
  geom_line() +
  scale_x_date(date_labels = "%b-%Y") + theme(legend.position = "none") +
  labs(title = "Junction 2", x = "Time", y = "Number of Vehicles")

junction_2_plot

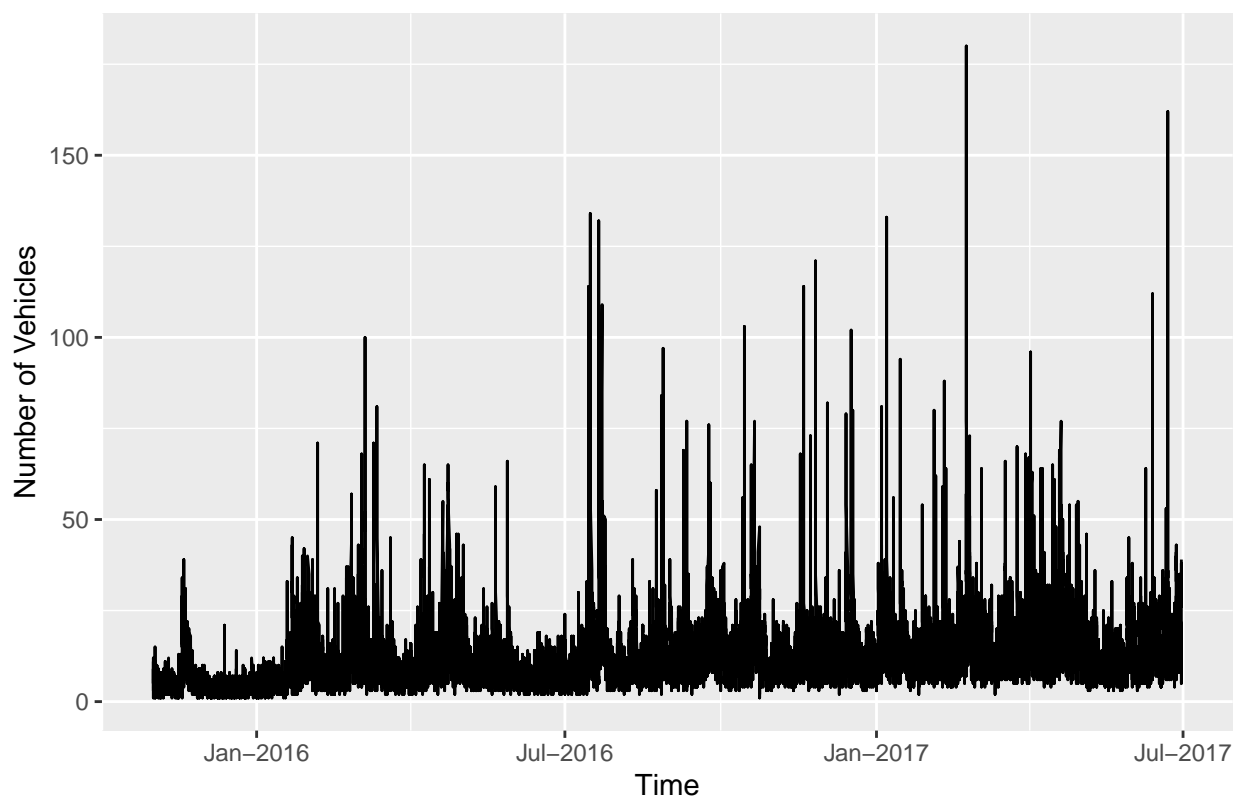
```

```
junction_3_plot <- ggplot(junctions_subset_3, aes(x = as.Date(junctions_subset_3$DateTime), y = Vehicles)) +
  geom_line() +
  scale_x_date(date_labels = "%b-%Y") + theme(legend.position = "none") +
  labs(title = "Junction 3", x = "Time", y = "Number of Vehicles")

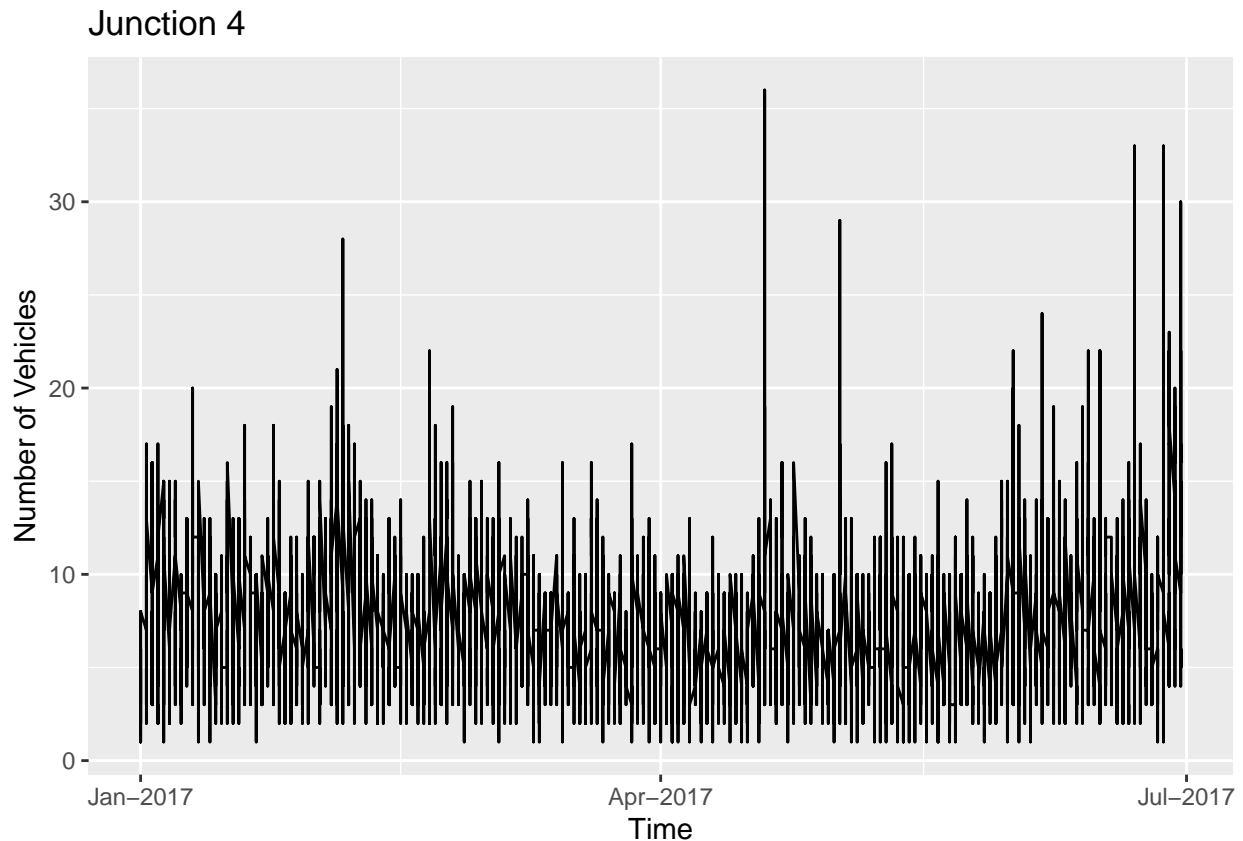
junction_3_plot
```

Junction 3



```
junction_4_plot <- ggplot(junctions_subset_4, aes(x = as.Date(junctions_subset_4$DateTime), y = VehicleCount)) +
  geom_line() +
  scale_x_date(date_labels = "%b-%Y") + theme(legend.position = "none") +
  labs(title = "Junction 4", x = "Time", y = "Number of Vehicles")

junction_4_plot
```



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

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

```
library(readxl)

alexa_data <- read_excel("/cloud/project/worksheet#4/Worksheet4c/alexa_file.xlsx")

num_obs <- nrow(alexa_data)
num_obs

## [1] 3150

num_cols <- ncol(alexa_data)
num_cols
```

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

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

```
var_counts <- alexa_data %>%
  count(variation)

var_counts
```

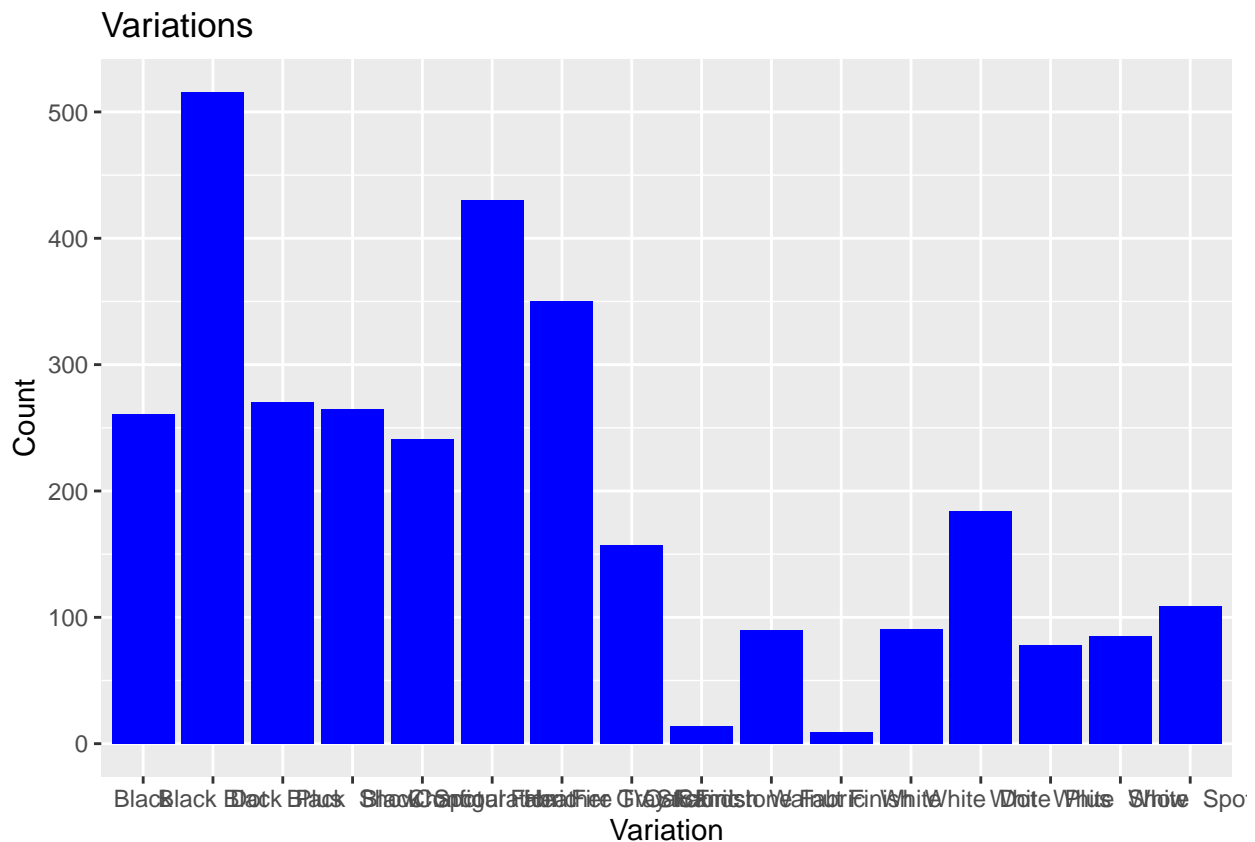
```
## # A tibble: 16 x 2
##   variation          n
##   <chr>          <int>
## 1 Black          261
```

##	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 graph. Show solution and answer.

```
alexa_plot <- ggplot(alexa_data, aes(x = variation)) +
  geom_bar(fill = "blue") +
  labs(title = "Variations",
       x = "Variation",
       y = "Count")
```

alexa_plot



The graph shows the distribution of variations and their respective counts. Each bar represents a spe

7d. Plot a `geom_line()` with the date and the number of verified reviews. Complete the details of the graphs. Show your answer and solution

```
library(dplyr)

alexa_data$date <- as.Date(alexa_data$date)

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

countMonth <- alexa_data %>%
  count(month)
countMonth

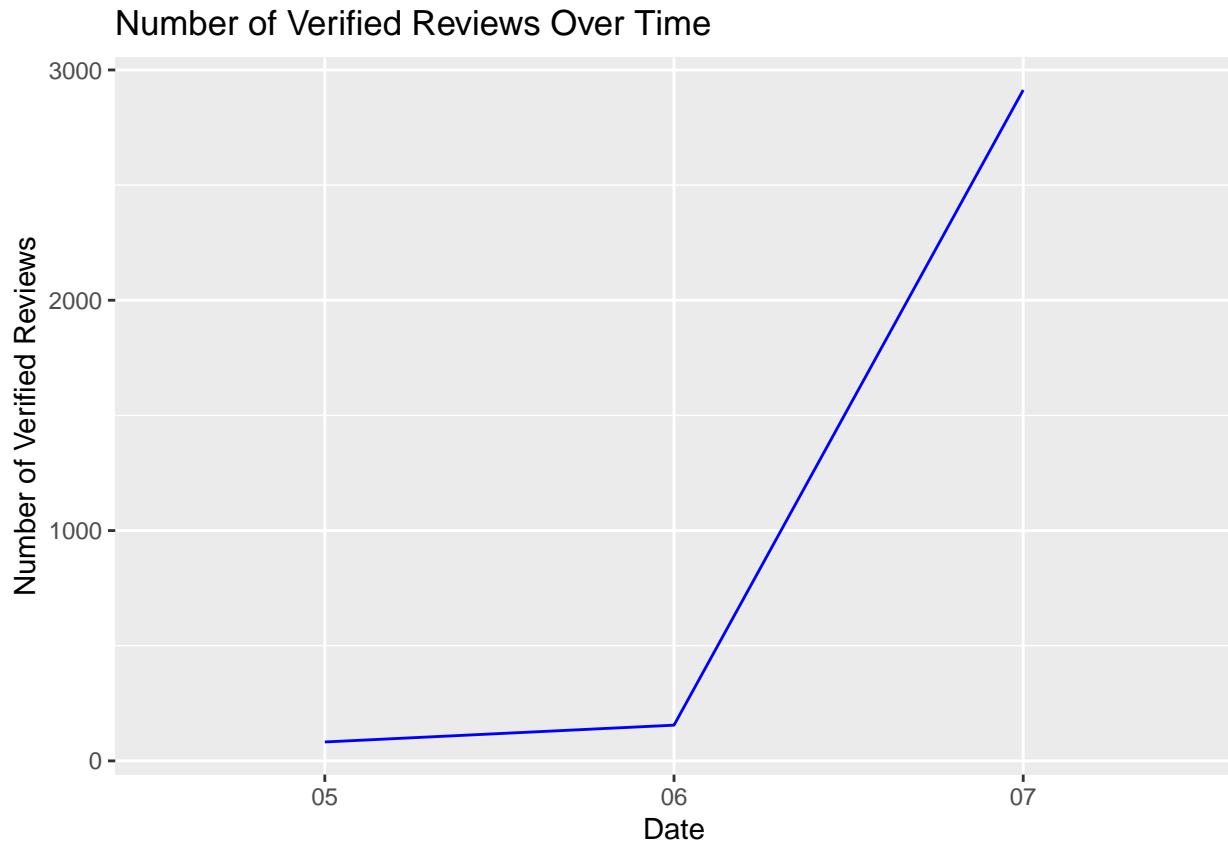
## # A tibble: 3 x 2
##   month     n
##   <chr> <int>
## 1 05      82
## 2 06     155
## 3 07    2913

monthly_revCount <- table(countMonth)
monthly_revCount

##           n
## month 82 155 2913
##    05  1   0   0
##    06  0   1   0
##    07  0   0   1

alexa_line <- ggplot(countMonth, aes(x = month, y = n, group = 1)) +
  geom_line(color = "blue") +
  labs(title = "Number of Verified Reviews Over Time",
       x = "Date",
       y = "Number of Verified Reviews")

alexa_line
```



7e. Get the relationship of variations and ratings. Which variations got the most highest in rating? Plot a graph to show its relationship. Show your solution and answer.

```
variation_ratings <- alexa_data %>%
  group_by(variation) %>%
  summarise(avg_rating = mean(rating))
```

```
variation_ratings
```

```
## # A tibble: 16 x 2
##   variation          avg_rating
##   <chr>             <dbl>
## 1 Black             4.23
## 2 Black Dot         4.45
## 3 Black Plus        4.37
## 4 Black Show        4.49
## 5 Black Spot        4.31
## 6 Charcoal Fabric   4.73
## 7 Configuration: Fire TV Stick 4.59
## 8 Heather Gray Fabric 4.69
## 9 Oak Finish        4.86
## 10 Sandstone Fabric  4.36
## 11 Walnut Finish     4.89
## 12 White             4.14
## 13 White Dot         4.42
## 14 White Plus        4.36
## 15 White Show        4.28
## 16 White Spot        4.31
```

```
highest_ratings <- variation_ratings %>%
  filter(avg_rating == max(avg_rating))
```

```
highest_ratings
```

```
## # A tibble: 1 x 2
##   variation    avg_rating
##   <chr>        <dbl>
## 1 Walnut Finish    4.89
```

```
# The walnut finish variation has the highest rating
```

```
ggplot(variation_ratings, aes(x = variation, y = avg_rating)) +
  geom_bar(stat = "identity", fill = "blue") +
  labs(title = "Average Ratings by Variation",
       x = "Variation",
       y = "Average Rating")
```

