# Data Visualiation in Tidyverse

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January 10, 2023

This R-markdown file uses tidyverse package in R to do data visualization exercise. In particular, it does the following:

#### Part I: We use earthquake data quakes of R to visualize it.

- 1. Plotting the distribution of earthquake magnitudes using tidyverse. In particular, we plot
  - i) histogram ii) boxplot iii) empirical cdf iv)Q-Q plot.
- 2. We plot earthquakes point on top of a map layer.

#### Part II: We use mpg dataset to use cool visualization functions of tidyverse

- 1. Aesthetic mapping of color
- 2. Different type of smoothing curves and colored points using group identity of a variable.
- 3. Using Facets function to add additional variable(s) to a 2D plot.
- 4. Playing with stat function.
- 5. Position adjustment options for 'geom\_bar()

#### Load necessary packages

```
# install the tidyverse package first if you have not done it yet.
#install.packages("tidyverse") # you can comment out this line after you have installed `tidyverse`
library(tidyverse) # for the `ggplot2` package
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr
             1.1.1
                       v readr
                                  2.1.4
## v forcats
           1.0.0
                                  1.5.0
                       v stringr
## v ggplot2 3.4.0
                       v tibble
                                  3.2.1
                       v tidyr
                                  1.3.0
## v lubridate 1.9.2
             1.0.1
## v purrr
## -- Conflicts ------tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                   masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become error
```

For rendering in PDF If you don't want to render (knit) the file in PDF format, you can ignore this block of code. If you face problem in rendering, please refer to debugging on https://yihui.org/tinytex/r/#debugging

.

```
#tinytex::reinstall_tinytex()
```

#### Main Analysis Starts From Here

**Dataset 1: Visualization the quakes data set in tidyverse** In our data visualization with Base R, we used the quakes data set contain the locations of 1000 seismic events of MB > 4.0. The events occurred in a cube near Fiji since 1964.

```
library(datasets)
?quakes
```

## starting httpd help server ... done

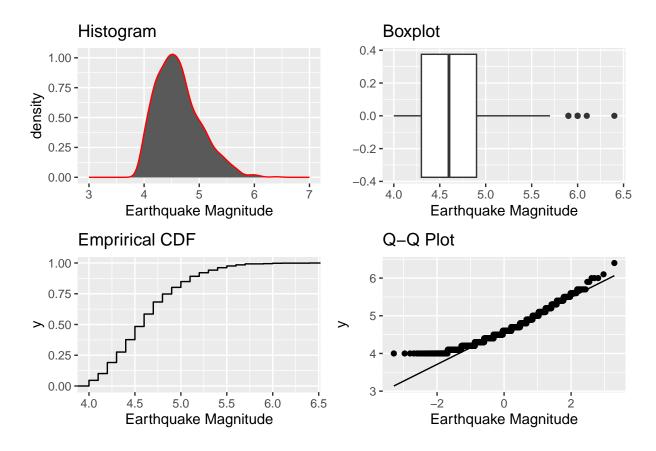
```
class(quakes)
## [1] "data.frame"
head(quakes, n=5) # print first 5 rows of quakes
             long depth mag stations
## 1 -20.42 181.62
                    562 4.8
                                  41
## 2 -20.62 181.03
                    650 4.2
                                  15
## 3 -26.00 184.10
                     42 5.4
                                  43
## 4 -17.97 181.66
                    626 4.1
                                  19
## 5 -20.42 181.96
                    649 4.0
                                  11
dim(quakes) # dimension of the table
## [1] 1000
              5
names(quakes) # list the variables in quakes
                 "long"
                          "depth"
                                      "mag"
## [1] "lat"
                                                  "stations"
str(quakes) # list the structures in quakes
## 'data.frame':
                 1000 obs. of 5 variables:
## $ lat
            : num -20.4 -20.6 -26 -18 -20.4 ...
             : num 182 181 184 182 182 ...
## $ long
## $ depth
            : int 562 650 42 626 649 195 82 194 211 622 ...
## $ mag
             : num 4.8 4.2 5.4 4.1 4 4 4.8 4.4 4.7 4.3 ...
## $ stations: int 41 15 43 19 11 12 43 15 35 19 ...
glimpse(quakes) # qet a qlimpse of the quakes data
## Rows: 1.000
## Columns: 5
## $ lat
             <dbl> -20.42, -20.62, -26.00, -17.97, -20.42, -19.68, -11.70, -28.1~
             <dbl> 181.62, 181.03, 184.10, 181.66, 181.96, 184.31, 166.10, 181.9~
## $ long
             <int> 562, 650, 42, 626, 649, 195, 82, 194, 211, 622, 583, 249, 554~
## $ depth
## $ mag
             <dbl> 4.8, 4.2, 5.4, 4.1, 4.0, 4.0, 4.8, 4.4, 4.7, 4.3, 4.4, 4.6, 4~
## $ stations <int> 41, 15, 43, 19, 11, 12, 43, 15, 35, 19, 13, 16, 19, 10, 94, 1~
```

#### a) Plotting the distribution of earthquake magnitudes

Writing ggplot2 code to reproduce the following four subfigures in a 2-by-2 layout.

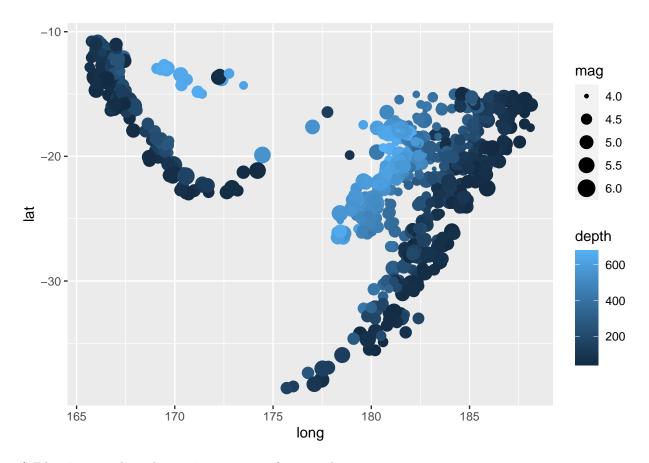
- subfigure #1: plot a density histogram of the earthquake magnitudes, and then plot the estimated probability density curve in red color in the same plot
- subfigure #2: plot a horizontal boxplot of the earthquake magnitudes
- subfigure #3: plot the empirical cdf of the earthquake magnitudes
- subfigure #4: make a Q-Q plot to compare the observed earthquake magnitudes distribution with the Normal distribution. Add a *thick* Q-Q line in blue color.

```
library(maps)
##
## Attaching package: 'maps'
## The following object is masked from 'package:purrr':
##
##
       map
library(datasets)
library(patchwork)
par(mfrow=c(2,2))
#subfigure-1
plot1=ggplot(data=quakes)+
geom_histogram(aes(mag), stat="density")+
xlab("Earthquake Magnitude")+
ggtitle("Histogram")+
geom_density(aes(mag), col="red")+
xlim(c(3,7))
#subfigure-2
plot2=ggplot(data=quakes)+
geom_boxplot(aes(mag))+
xlab("Earthquake Magnitude")+
ggtitle("Boxplot")
#subfigure-3
plot3=ggplot(data=quakes)+
stat ecdf(aes(mag))+
xlab("Earthquake Magnitude")+
ggtitle("Emprirical CDF")
#subfigure-4
plot4=ggplot(data=quakes,mapping=aes(sample=mag))+
geom_qq()+ geom_qq_line()+
xlab("Earthquake Magnitude")+
ggtitle("Q-Q Plot")
plot1+plot2+plot3+plot4
```



b) Earthquake location map Scatter plot of the earthquake locations. Use long as the x-axis and lat as the y-axis. Map mag to the size aesthetic and depth to the color aesthetic.

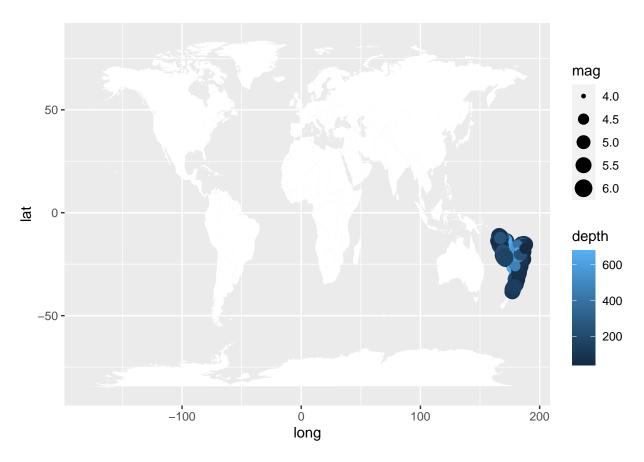
```
ggplot(data=quakes, mapping=aes(x = long, y = lat, colour=depth, size=mag))+
geom_point()
```



## c) Plotting earthquakes point on top of a map layer

```
library(maps)
wc=map_data("world")

ggplot()+
geom_map(data=wc, map=wc, aes(long,lat, map_id=region), fill="white")+
geom_point(data=quakes, mapping=aes(x = long, y = lat, colour=depth, size=mag))
```



Dataset 2: Visualization of the mpg data set This data set contains fuel economy data 1999 - 2008 for 38 popular car models (source: EPA http://fueleconomy.gov).

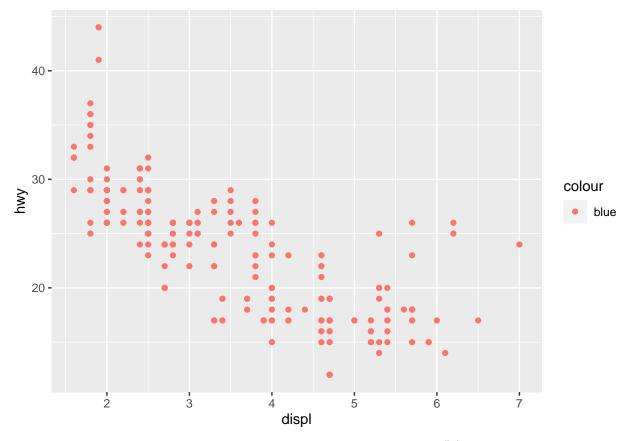
```
?mpg
dim(mpg) # dimension of the table
## [1] 234 11
names(mpg) # list the variables in mpg
   [1] "manufacturer" "model"
                                                    "year"
                                                                   "cyl"
  [6] "trans"
                       "drv"
                                     "cty"
                                                    "hwy"
                                                                   "fl"
## [11] "class"
str(mpg) # list the structures in mpg
## tibble [234 x 11] (S3: tbl df/tbl/data.frame)
## $ manufacturer: chr [1:234] "audi" "audi" "audi" "audi" ...
   $ model
##
                : chr [1:234] "a4" "a4" "a4" "a4" ...
## $ displ
                : num [1:234] 1.8 1.8 2 2 2.8 2.8 3.1 1.8 1.8 2 ...
                : int [1:234] 1999 1999 2008 2008 1999 1999 2008 1999 1999 2008 ...
## $ year
                 : int [1:234] 4 4 4 4 6 6 6 4 4 4 ...
## $ cyl
                 : chr [1:234] "auto(15)" "manual(m5)" "manual(m6)" "auto(av)" ...
##
   $ trans
                 : chr [1:234] "f" "f" "f" "f" ...
## $ drv
## $ cty
                 : int [1:234] 18 21 20 21 16 18 18 18 16 20 ...
                 : int [1:234] 29 29 31 30 26 26 27 26 25 28 ...
##
   $ hwy
                 : chr [1:234] "p" "p" "p" "p" ...
##
   $ fl
                : chr [1:234] "compact" "compact" "compact" "...
glimpse(mpg) # get a glimpse of the mpg data
## Rows: 234
## Columns: 11
```

```
## $ manufacturer <chr> "audi", "audi"
                                                          <chr> "a4", "a4", "a4", "a4", "a4", "a4", "a4", "a4", "a4 quattro", "~
## $ model
## $ displ
                                                          <dbl> 1.8, 1.8, 2.0, 2.0, 2.8, 2.8, 3.1, 1.8, 1.8, 2.0, 2.0, 2.~
                                                          <int> 1999, 1999, 2008, 2008, 1999, 1999, 2008, 1999, 1999, 200~
## $ year
## $ cyl
                                                           <int> 4, 4, 4, 4, 6, 6, 6, 4, 4, 4, 4, 6, 6, 6, 6, 6, 6, 8, 8, ~
## $ trans
                                                           <chr> "auto(15)", "manual(m5)", "manual(m6)", "auto(av)", "auto~
                                                           ## $ drv
                                                           <int> 18, 21, 20, 21, 16, 18, 18, 18, 16, 20, 19, 15, 17, 17, 1~
## $ cty
                                                           <int> 29, 29, 31, 30, 26, 26, 27, 26, 25, 28, 27, 25, 25, 25, 2~
## $ hwy
                                                           ## $ fl
## $ class
                                                           <chr> "compact", "compact", "compact", "compact", "c~
```

#### a) Aesthetic mapping of color

i) The following codes does not show points in blue color. What's gone wrong with the following code? Why are the points not blue?

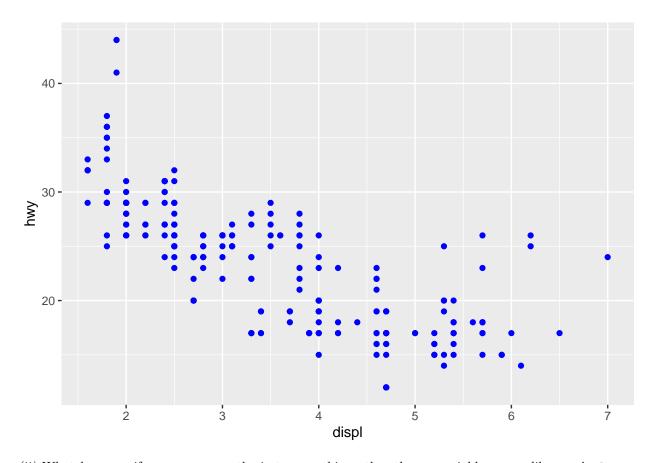
```
ggplot(data = mpg) +
geom_point(mapping = aes(x = displ, y = hwy, color = "blue"))
```



Answer: Points were not blue because color="blue" was written inside aes()/mapping which is not read as argument color by R, it is read as a vector c("blue") to map to an aesthetic, just like x and y variables displ and hwy. After writing color outside the mapping/aes, the R reads it as  $color^{**}$ .

Task: Correct the code to plot blue points.

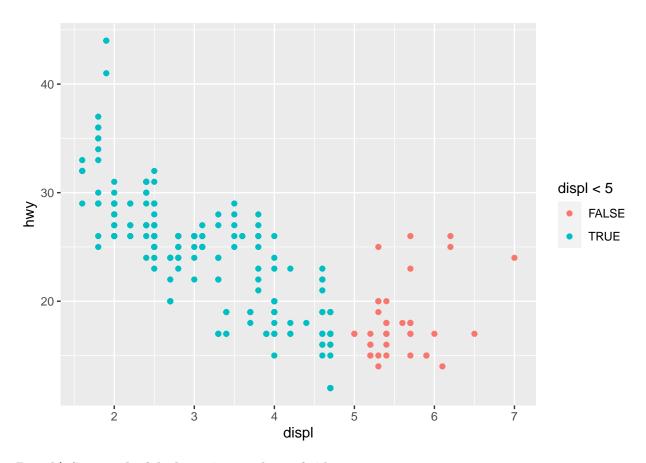
```
ggplot(data = mpg) +
geom_point(aes(x = displ, y = hwy), color = "blue")
```



(ii) What happens if we map an aesthetic to something other than a variable name, like aes(colour = displ < 5)?

Answer If we map an aesthetic to something other than a variable name, like 'aes(colour= displ<5) then ggplot() function works like a temporary variable is added in the data with values equal to the result of expression. In our case, it takes values of 'TRUE' Or 'FALSE' because displ<5 is a logical variable.x and y are displ and hwy as before.

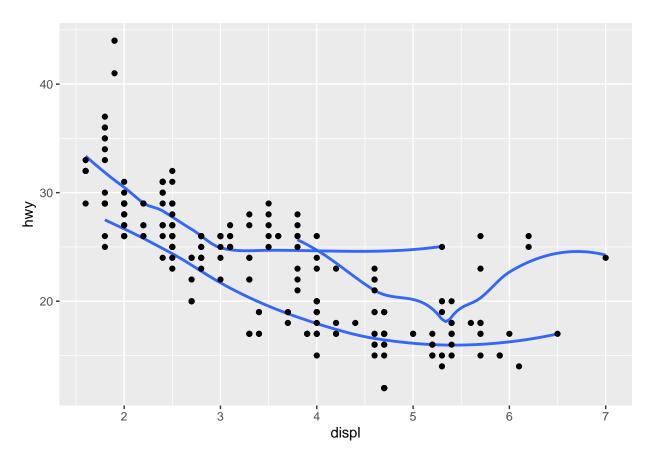
```
ggplot(mpg, aes(x = displ, y = hwy, colour = displ < 5)) +
  geom_point()</pre>
```



Part b) Some colorful plot using ggplot2 of tidyverse

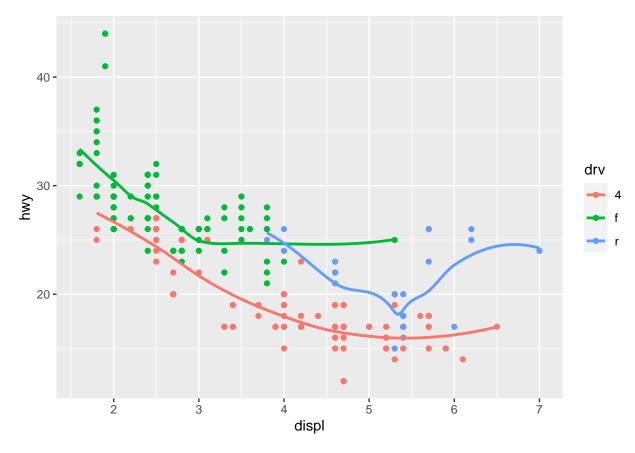
i) Smooth plots according to group identity but same color. Group is  $\mathtt{drv}$  variable which is a categorical variable taking 3 values:  $\{\mathtt{r},\mathtt{f},\mathtt{4}\}$ . Where  $\mathtt{r}$  is for rear wheels,  $\mathtt{f}$  is for front wheels, and 4 stands for all four wheels.

```
# Enter your code here
ggplot(mpg, aes(x = displ, y = hwy)) +
  geom_smooth(mapping = aes(group = drv), se = FALSE) +
  geom_point()
## `geom_smooth()` using method = 'loess' and formula = 'y ~ x'
```



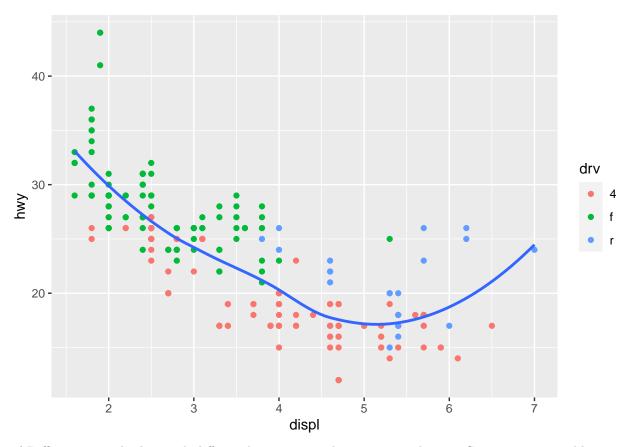
ii) Smooth plots and colors according to group identity. Group is drv variable which is a categorical variable taking 3 values:  $\{r,f,4\}$ . WHERE r IS for rear wheels, f is for front wheels, and 4 stands for all four wheels.

```
ggplot(mpg, aes(x = displ, y = hwy, colour = drv)) +
  geom_point() +
  geom_smooth(se = FALSE)
## `geom_smooth()` using method = 'loess' and formula = 'y ~ x'
```



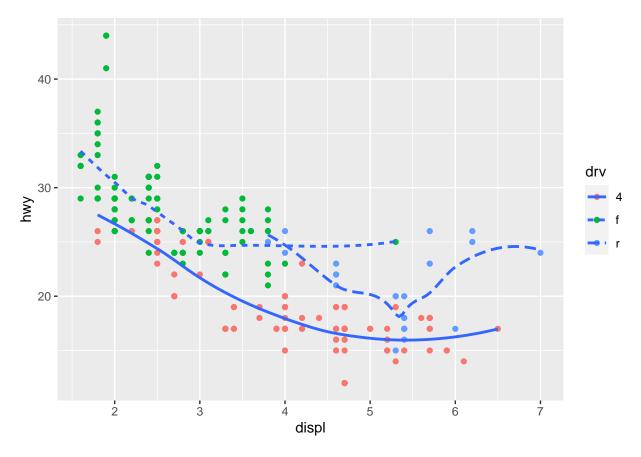
iii) Single smooth plot but data points being colored according to group identity. Group is drv variable which is a categorical variable taking 3 values:  $\{r,f,4\}$ . WHERE r IS for rear wheels,f is for front wheels, and 4 stands for all four wheels.

```
ggplot(mpg, aes(x = displ, y = hwy)) +
  geom_point(aes(colour = drv)) +
  geom_smooth(se = FALSE)
## `geom_smooth()` using method = 'loess' and formula = 'y ~ x'
```



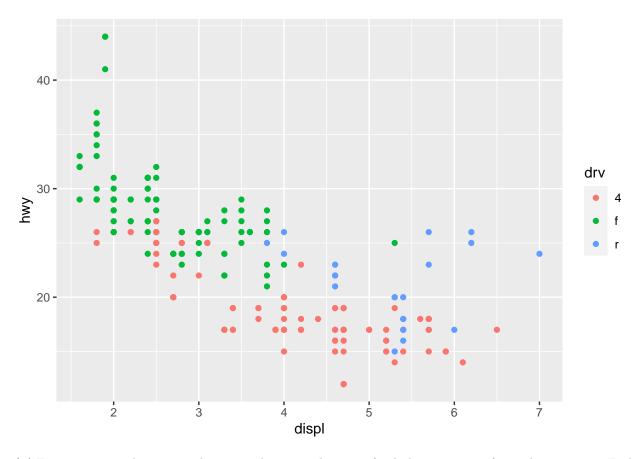
iv) Diffierent smooth plots with different linetype according to group identity. Group is drv variable.

```
ggplot(mpg, aes(x = displ, y = hwy)) +
geom_point(aes(colour = drv)) +
geom_smooth(aes(linetype = drv), se = FALSE)
## `geom_smooth()` using method = 'loess' and formula = 'y ~ x'
```



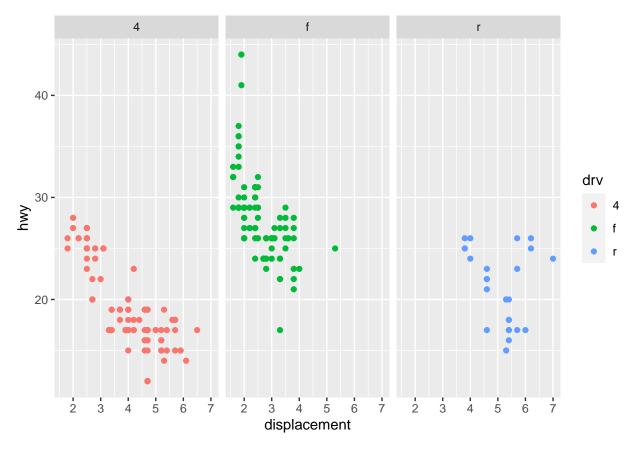
- c) Facets: to add additional variable(s) to a 2D plot There are two ways to add additional variable(s) to a 2D plot. One is using aesthetics, the other one is using facets.
- (i) Scatter plot: x-axis is displ and y-axis is hwy. We use different colors to distinguish drv types.

```
ggplot(data=mpg, mapping=aes(x = displ, y = hwy, colour=drv))+
  geom_point()
```



(ii) Facet drv into the rows. That is, makes several **rows** of subplots, one row for each drv type. Each subplot has displ mapped to the x-axis and hwy mapped to the y-axis. [Note: Use nrow or ncol to control the layout of the individual panels].

```
ggplot(data=mpg, aes(x=displ, y=hwy, color=drv))+
geom_point()+
labs(x="displacement", y="hwy")+
scale_color_discrete(name="drv")+
facet_wrap(~drv, nrow=1)
```



### d) stat functions

Most geom functions and stat functions come in pairs that are almost always used in concert.

- every geom has a default stat
- · every stat has a default geom

Look up the default stat functions for the geom functions listed in the following table. The variables computed by the default stat function (Reference: the Computed variables section in the R-documentation page).

geom function	default stat function	variables computed by the default stat function
geom_bar()	count	Frequency
${\tt geom\_histogram}()$	bin	Frequency
<pre>geom_density()</pre>	density	Density, Count
<pre>geom_point()</pre>	identity	Sum of values
<pre>geom_smooth()</pre>	smooth	Smoothed or locally averaged value

Question: Some geom function has stat = "identity" as the default. What does that mean?

Answer: If there are 3 teams A,B, and c with equal occurrence. Then geom\_bar function will create bar chart displaying the count of occurrence in the games, which is equal. But, If we use stat="identity" with geom\_bar then bar chart will be created displaying sum of points scored by the teams in each game. (Additional Note: Table formatting are sometimes tricky using R Markdown. Table Generator is a handy tool if you need to make tables in the future.)

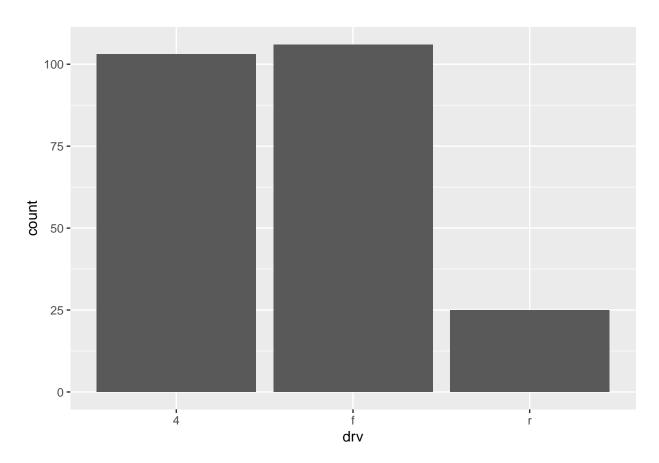
#### e) Position adjustment options for geom\_bar()

Using two categorical variables from the mpg data set and to illustrate the following four position adjustment options for geom\_bar():

- **default**: position = "stack"
- position = "identity" will place each object exactly where it falls in the context of the graph.
- position = "fill" works like stacking, but makes each set of stacked bars the same height.
- position = "dodge" places overlapping objects directly beside one another, the bars are automatically stacked. Each colored rectangle represents a combination of cut and clarity.

```
i) position = "stack"
```

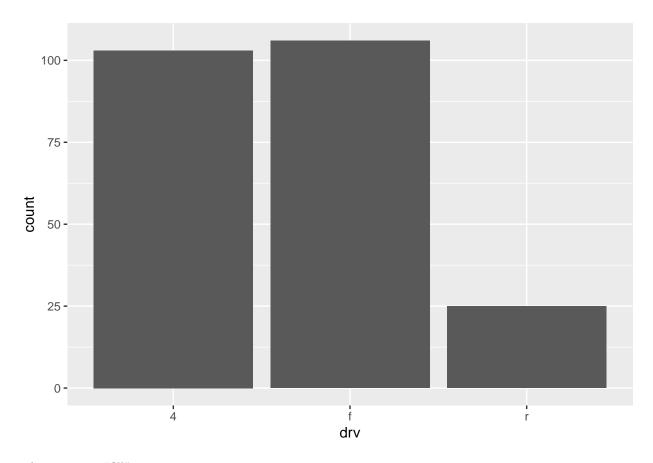
```
ggplot(data=mpg)+
geom_bar( mapping= aes(drv) , position = "stack")
```



ii) position = "identity"

It will place each object exactly where it falls in the context of the graph.

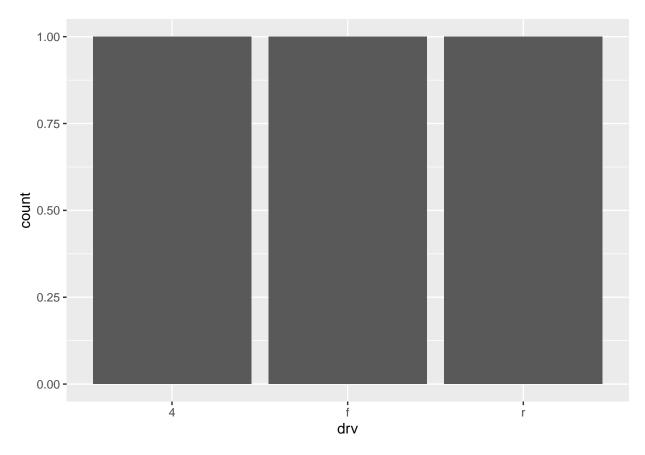
```
ggplot(data=mpg)+
geom_bar( mapping= aes(drv) , position = "identity")
```



iii) position = "fill"

It works like stacking, but makes each set of stacked bars the same height.

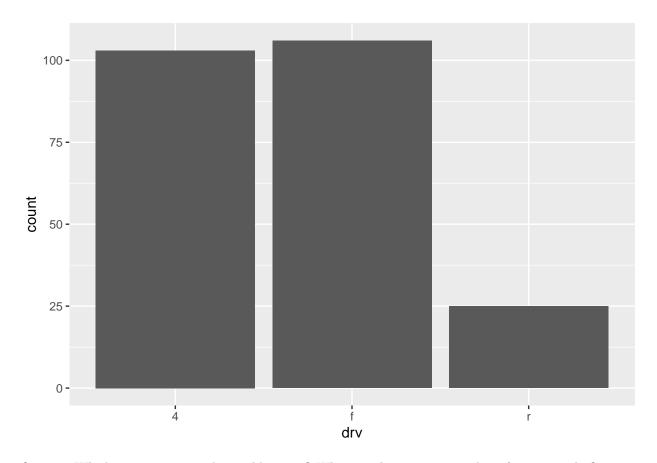
```
ggplot(data=mpg)+
geom_bar( mapping= aes(drv) , position = "fill")
```



iv) position = "dodge"

dodge places overlapping objects directly beside one another.

```
ggplot(data=mpg)+
geom_bar( mapping= aes(drv) , position = "dodge")
```



Question Which position option do you like most? What conclusions can you draw from your plot?

ANSWER: I liked position = "dodge" which places overlapping objects directly beside one another. Choosing right position argument is important part of making good plot. You may like other position, it is not really a right or wrong answer.

## Acknowledgments

List of all the help I have received for completing this work. 1. I used rdocumentation.org website to learn more about the functions. 2. I used geeksforgeeks.com website to get help on tidyverse paradigm of R.