



ALVAREZ

College of Business

The University of Texas at San Antonio

Introduction to Programming in R

Module 3: Communication

Learning Objectives

- Communicating with Markdown
- Creating .Rmd files
- Output and knitting .Rmd
- Visualization with ggplot2
- Creating a plot with ggplot2
- Scatterplots, bar charts, histograms, and other plots
- Aesthetics, facets, plotting objects, and trendlines
- Overlaying multiple plots

Communicating

- R markdown lets you create documents that can help you communicate your findings to decision-makers (including yourself).
- Elegant and concise way to record and present your code, results, and thoughts.
- Makes research reproducible, so that others can easily understand and replicate, if needed.
- I find this particularly helpful working with graduate students. They can show me what they did, exactly, and provide narration with it.
- There are some basic Markdown syntax that we'll cover here. As you progress with this program and in your career, you'll use more.
- Markdown can output into .pdf, .doc, .html, and other popular file formats.
- Also cover some plotting techniques to produce elegant visuals. There will be more on visuals in the Visualization course.

R Markdown Basics

- Markdown file (.Rmd) can help with:
 - Communicating with decision-makers with focus on insights and takeaways, rather than code.
 - Collaborating with other researchers, as well as future you, with focus of understanding the code and the results.
 - Notetaking for other users, as well as future you, with a focus of documenting your thought process.
- Merges many external packages to work. So the built-in help function ? is limited.
 - R Markdown Cheat Sheet: *Help > Cheatsheets > R Markdown Cheat Sheet*,
 - R Markdown Reference Guide: *Help > Cheatsheets > R Markdown Reference Guide*.
- Markdown uses “rmarkdown” package, but R-studio comes with it loaded.

What .Rmd looks like

- Typical markdown looks something like this... with code and prose mixed.

```
---
title: "Creating A New Markdown"
author: "Arka Roy"
date: "1/22/2020"
output: html_document
---
```

*Yet Another Markup Language (YAML) header
with --- above and below.*

```
```{r setup, include=FALSE}
knitr::opts_chunk$set(echo = TRUE)
```
```

Chunks of R-code, with ``` before and after.

```
## R Markdown
```

This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more details on using R Markdown see <<http://rmarkdown.rstudio.com>>.

When you click the **Knit** button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

```
```{r cars}
summary(cars)
```
```

*Texts with simple formatting,
for e.g., # header or _word_ for italics.*

```
## Including Plots
```

You can also embed plots, for example:

```
```{r pressure, echo=FALSE}
plot(pressure)
```
```

Note that the `echo = FALSE` parameter was added to the code chunk to prevent printing of the R code that generated the plot.

Creating and Running .Rmd

```

```{r setup, include=FALSE}
knitr::opts_chunk$set(echo = TRUE)
```

## R Markdown

This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more details on using R Markdown see
http://rmarkdown.rstudio.com.

When you click the Knit button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

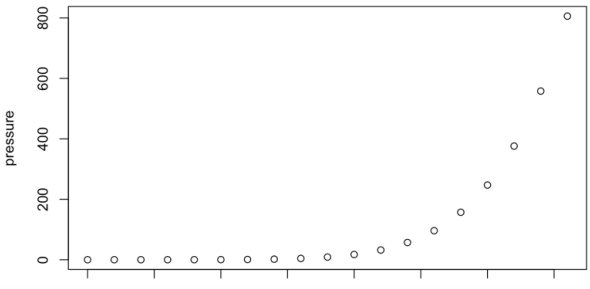
```{r cars}
summary(cars)
```

  speed      dist
  Min.   : 4.0   Min.   : 2.00
  1st Qu.:12.0   1st Qu.: 26.00
  Median :15.0   Median : 36.00
  Mean   :15.4   Mean   : 42.98
  3rd Qu.:19.0   3rd Qu.: 56.00
  Max.   :25.0   Max.   :120.00

## Including Plots

You can also embed plots, for example:

```{r pressure, echo=FALSE}
plot(pressure)
```



```

- Create a .Rmd by following the wizard at:

File > New File > R Markdown

- Can run a full code chunks one-at-a-time using *cmd/ctrl+shift+return* or

>



- Results are displayed inline with text.

- A report of the full writeup and results can be knitted by *cmd/ctrl+shift+K* or



- The report is made in .html by default, but can be also done as .pdf, .doc, etc.

Text formatting in .Rmd

```
Text formatting
-----

*italic* or _italic_
**bold**  __bold__
`code`
superscript^2^ and subscript~2~

Headings
-----

# 1st Level Header
## 2nd Level Header
### 3rd Level Header

Lists
-----

* Bulleted list item 1
* Item 2
  | * Item 2a
  | * Item 2b
1. Numbered list item 1
1. Item 2. The numbers are incremented automatically in the output.

Tables
-----

First Header	Second Header
Content Cell | Content Cell
Content Cell | Content Cell
```



Text formatting

italic or *italic* **bold** **bold** `code` superscript² and subscript₂

Headings

1st Level Header

2nd Level Header

3rd Level Header

Lists

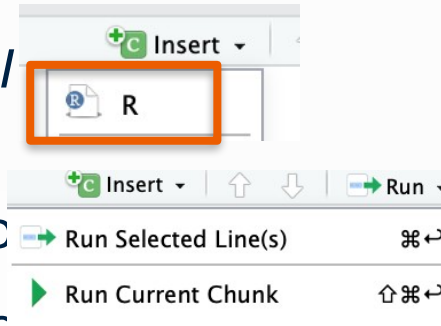
- Bulleted list item 1
- Item 2
 - Item 2a
 - Item 2b
- 1. Numbered list item 1
- 2. Item 2. The numbers are incremented automatically in the output.

Tables

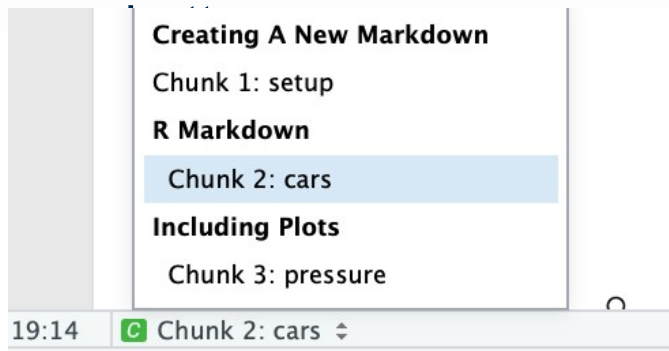
| First Header | Second Header |
|--------------|---------------|
| Content Cell | Content Cell |
| Content Cell | Content Cell |

Code Chunks in .Rmd

- Insert a code chunk in markdown by *cmd/ctrl+shift+I* typing in ```{r}` and ````` before and after code.
- You can run a code chunk by *cmd/ctrl+shift+return* or manually
- Using *cmd/ctrl+return* will still only run the single line of code.
- Inside the ```{r}` and `}` you can define your own name for the chunk. For example, the 2nd chunk is named cars and it can be easily accessed via the navigation pane at the



```
``{r cars}
summary(cars)
```
```




Option	Execute code	Show code	Output	Plots	Messages	Warnings
eval=false		✓				
include=false	✓					
echo=false	✓		✓	✓	✓	✓
results='hide'	✓	✓		✓	✓	✓
fig.show='hide'	✓	✓	✓		✓	✓
message=false	✓	✓	✓	✓		✓
warning=false	✓	✓	✓	✓	✓	

- Chunks have optional arguments:



# Output from .Rmd

- Common file format outputs:
  - .pdf: need access to a LaTeX compiler, such as MikTeX or TexWorks.
  - .doc: helpful for reports in Microsoft Word.
  - .rft: rich text format can also be used.
  - github\_document: helpful for sharing code, analysis, & prose on the net.
- You can set global options for the .Rmd file:

```
```{r setup, include=FALSE}  
knitr::opts_chunk$set(echo = TRUE)  
```
```
- When sharing with decision-makers, use *echo=FALSE*.
- For sharing *.html*, you can hide code by default, and make it accessible by a mouse click. Use *code\_folding: hide* in header.
- Can also do inline code- e.g. `The total of v1 is `r sum(v1)``  `The total of v1 is 55`

# Other .Rmd outputs

- R Notebook - like a lab notebook, just for R tasks.
  - .html document: communicate with decision-makers
  - .html notebook: collaborate with data scientists and yourself.
  - Unlike .html document, .html notebook will show full source code.
  - Can be viewed in browser or R-studio.
- Presentations – less visual control than powerpoint but easier to transfer data science commands.
  - New slides starts at # header or ## subheader
  - Horizontal line with \*\*\*
  - Can make two types of html presentations or pdf presentation via beamer.

# Tidyverse

- Will use the *tidyverse* package throughout the next few modules.
  - Use `install.packages("tidyverse")` ≈ first time only
  - Load by `library(tidyverse)` ≈ everytime

```
— Conflicts —
x dplyr::filter() masks stats::filter()
x dplyr::lag() masks stats::lag()
... .
```

- What does this conflict mean?
  - Base R has a stats package where it has the functions *filter()* and *lag()*.
  - The *tidyverse* package has the same name functions. So it overwrites it.
  - Not to worry, the *tidyverse* versions do the same stuff and more...

# ggplot2 package

- R has many plotting functions and packages, but ggplot2 package is very elegant yet powerful.
- It is built into the tidyverse package, so you will need to install it or load it separately.
- Throughout this module let us use another mileage data that's built into ggplot2 package. To load: *data("mpg")* and *print(mpg)*.

```
> print(mpg)
```

```
A tibble: 234 x 11
```

|    | manufacturer | model      | displ | year  | cyl   | trans      | drv   | cty   | hwy   | fl    | class   |
|----|--------------|------------|-------|-------|-------|------------|-------|-------|-------|-------|---------|
|    | <chr>        | <chr>      | <dbl> | <int> | <int> | <chr>      | <chr> | <int> | <int> | <chr> | <chr>   |
| 1  | audi         | a4         | 1.8   | 1999  | 4     | auto(l5)   | f     | 18    | 29    | p     | compact |
| 2  | audi         | a4         | 1.8   | 1999  | 4     | manual(m5) | f     | 21    | 29    | p     | compact |
| 3  | audi         | a4         | 2     | 2008  | 4     | manual(m6) | f     | 20    | 31    | p     | compact |
| 4  | audi         | a4         | 2     | 2008  | 4     | auto(av)   | f     | 21    | 30    | p     | compact |
| 5  | audi         | a4         | 2.8   | 1999  | 6     | auto(l5)   | f     | 16    | 26    | p     | compact |
| 6  | audi         | a4         | 2.8   | 1999  | 6     | manual(m5) | f     | 18    | 26    | p     | compact |
| 7  | audi         | a4         | 3.1   | 2008  | 6     | auto(av)   | f     | 18    | 27    | p     | compact |
| 8  | audi         | a4 quattro | 1.8   | 1999  | 4     | manual(m5) | 4     | 18    | 26    | p     | compact |
| 9  | audi         | a4 quattro | 1.8   | 1999  | 4     | auto(l5)   | 4     | 16    | 25    | p     | compact |
| 10 | audi         | a4 quattro | 2     | 2008  | 4     | manual(m6) | 4     | 20    | 28    | p     | compact |

```
... with 224 more rows
```

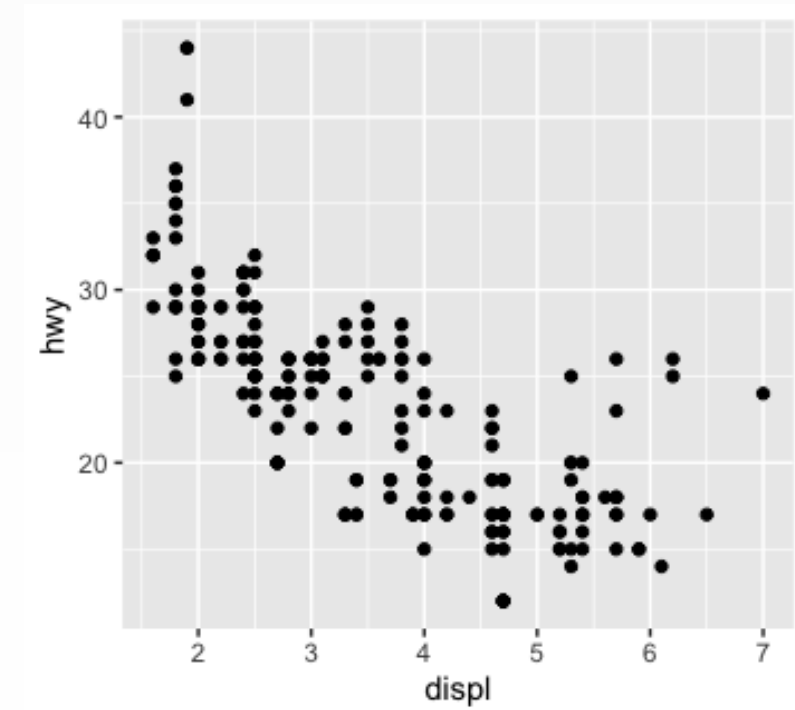
?mpg to find out  
more details about  
variables.

# Creating a *ggplot*

- Question: Do cars with bigger engines use more gas than cars with smaller engines?
- Let us plot hwy (miles per gallon on the highway) against displ (car's engine size, in liters).

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

- As expected, we see a negative relationship.
- Cars with bigger engines have lower mpg.



# *ggplot()*

- To plot with ggplot2, you start with the function *ggplot()*, which creates a coordinate system for the plot.
  - If you try this, it'll just be a blank canvas.
- Then, the first argument is the data. Here, we used the mpg data.
  - Even when you add the data argument, the plot remains blank.

```
> ggplot(data = mpg) +
 + geom_point(mapping = aes(x = displ, y = hwy))
```
- Then, you add layers to the plot using +.
  - Note that the + must be added to the end of the line if the code is being broken into multiple lines.
- Then *geom* functions controls the type of plot. Many geom functions!
  - Here, *geom\_point()* adds a layer of points making it a scatterplot.
  - *Geom* functions take mapping arguments that defines how to map the data to visuals.

## *ggplot() (cont.)*

- The basic template of *ggplot()* is

```
ggplot(data = DATA) +
 <GEOM_FUNCTION>(mapping = aes(MAPPINGS))
```

- The mapping argument uses the aesthetic function *aes()*, and the *x* and *y* arguments of *aes()* specifies which variable goes on x-axis and y-axis.

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

- In our first example, we plotted the engine size on the x-axis and mileage on the y-axis.
- You can also add a 3<sup>rd</sup> factor variable to a 2D scatterplot by using *aes()*.

- For e.g., Color the observations by class of cars

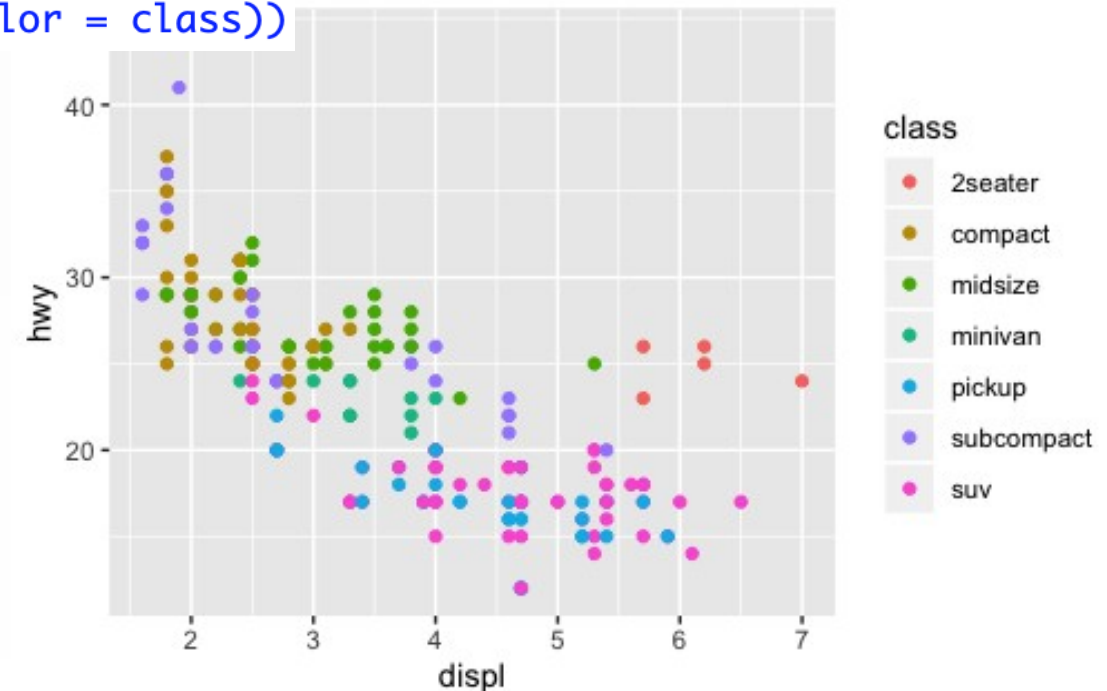
```
> levels(as.factor(mpg$class))
[1] "2seater" "compact" "midsize" "minivan" "pickup" "subcompact" "suv"
> unique(mpg$class)
[1] "compact" "midsize" "suv" "2seater" "minivan" "pickup" "subcompact"
```

# Aesthetics

- `aes()` includes visual properties like size, shape, or color of points.
- So to color the observations by class of car -

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

- `ggplot()` will assign unique properties to the arguments level of the factor variable.
- For e.g. different colors are assigned to the 7 levels of class variable.
- `ggplot()` will also add a legend.



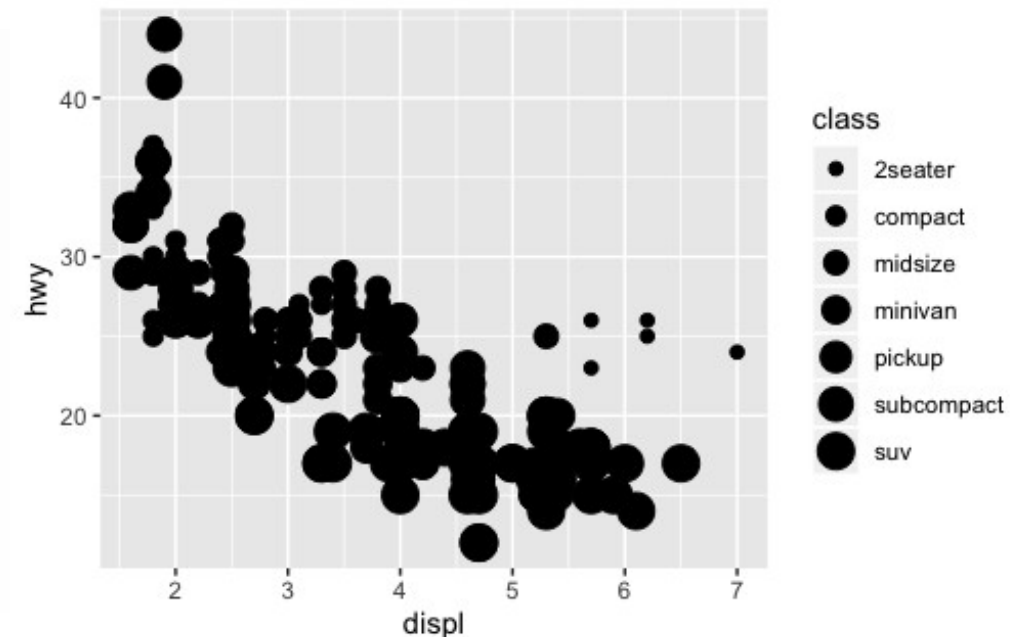


# Aesthetics (cont.)

- You can also change size of the points

```
> ggplot(data = mpg) +
+ geom_point(mapping = aes(x = displ, y = hwy, size = class))
Warning message:
Using size for a discrete variable is not advised.
```

- The warning here shows that we really should not map an unordered variable to an ordered aesthetic.
- Instead, we could use the number of cylinders, for example.



# Aesthetics (cont..)

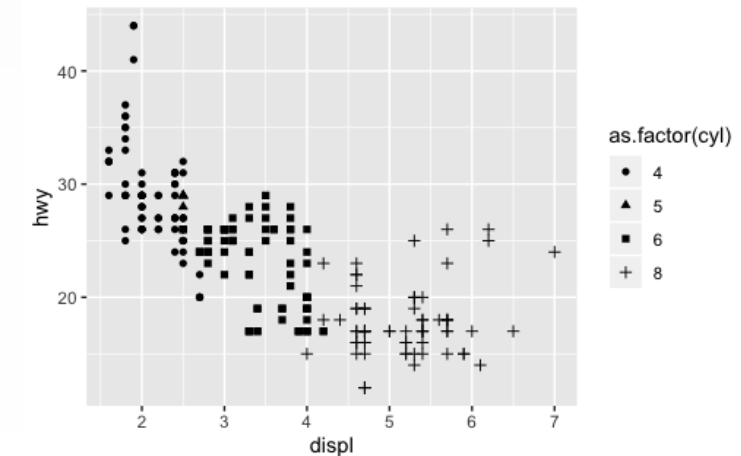
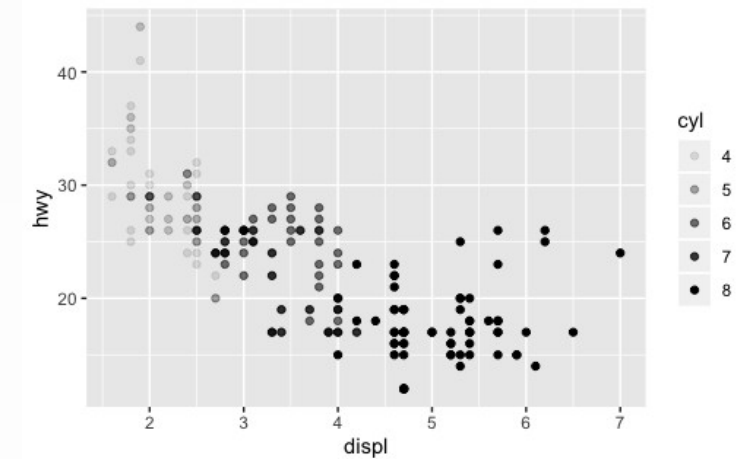
- `aes()` also has an *alpha* argument that maps to opacity of the points.

```
> ggplot(data = mpg) +
+ geom_point(mapping = aes(x = displ, y = hwy, alpha = class))
Warning message:
Using alpha for a discrete variable is not advised.
```

- Like *size*, it is better to use ordered numbers for *alpha*, so I used *cyl*.
- The argument *shape* controls the type of symbols used as points.

```
> ggplot(data = mpg) +
+ geom_point(mapping = aes(x = displ, y = hwy, shape = as.factor(cyl)))
```

- `ggplot()` only can plot 6 shapes at a time.
  - Additional groups will not be plotted. You can see this if you do *shape = class*.
- *shape* only accepts categorical variables.
  - You'll see the error if you don't use *as.factor()*.



# Aesthetics (cont...)

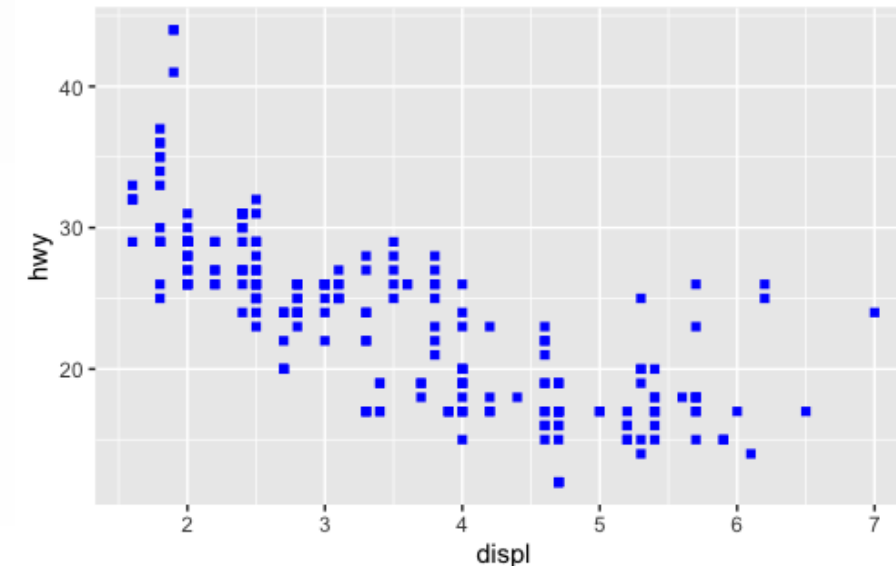
- You can also change the aesthetics of the entire *geom* (plot), manually.
- To do so, place the aesthetic argument outside of *aes()*.
- For e.g. if you wanted all the points to be blue squares.

```
> ggplot(data = mpg) +
+ geom_point(mapping = aes(x = displ, y = hwy), color = "blue", shape = 15)
```

Note that

- Name of color is a character string.
- Size of the symbol is in mm.
- Shape of symbol is a number.

|     |     |      |      |      |
|-----|-----|------|------|------|
| □ 0 | ✕ 4 | ⊕ 10 | ■ 15 | ■ 22 |
| ○ 1 | ▽ 6 | ⊗ 11 | ● 16 | ● 21 |
| △ 2 | ⊠ 7 | ⊞ 12 | ▲ 17 | ▲ 24 |
| ◇ 5 | ✱ 8 | ⊗ 13 | ◆ 18 | ◆ 23 |
| ⊥ 3 | ◊ 9 | ◻ 14 | ● 19 | ● 20 |

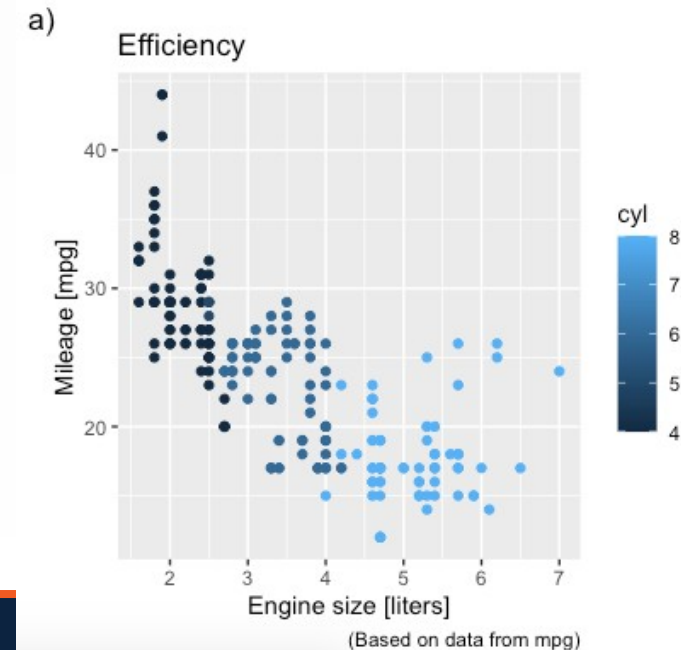


# Aesthetics (cont....)

- The *labs()* function allows you to change the default labels given by *ggplot2*.
- Use *labs(x = "string", y = "string")* to assigned new axis labels.

```
> ggplot(data = mpg) +
+ geom_point(mapping = aes(x = displ, y = hwy, color = cyl)) +
+ labs(x = "Engine size [liters]", y = "Mileage [mpg]", title = "Efficiency", caption = "(Based on data from mpg)", tag = "a)")
```

Arguments like *title*, *caption*, and *tag* are all useful for making descriptive plots.

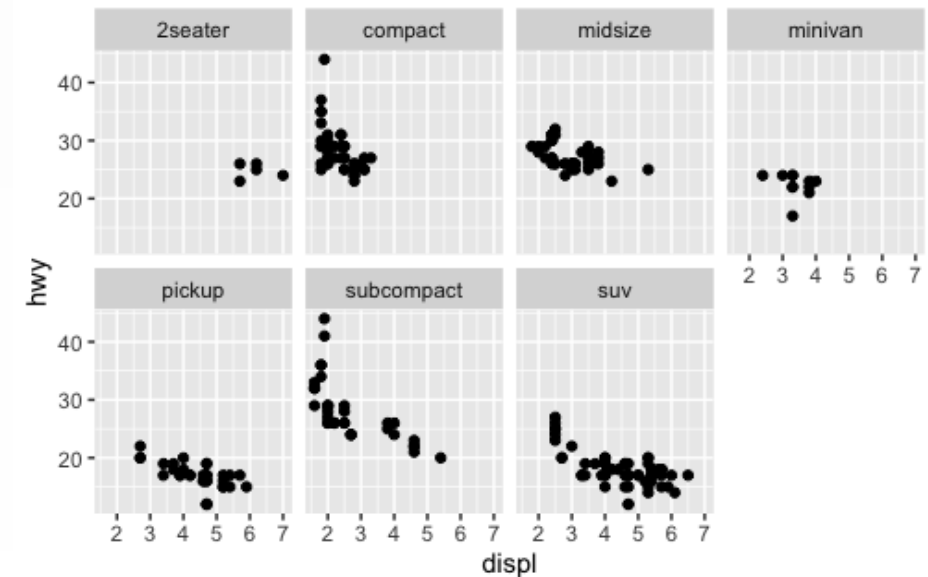


# Facets

- We can also make subplots within a plot - like a matrix of subplots.
- To make subplot based on using a factor variable use *facet\_wrap()*.
- The first argument is a formula – this is a data structure in R – not an “equation”.
  - A formula structure is created using a tilde followed by a variable.

```
> ggplot(data = mpg) +
+ geom_point(mapping = aes(x = displ, y = hwy)) +
+ facet_wrap(~ class, nrow = 2)
```

- You can specify the number of rows using *nrow* argument.
- The variable to pass to *facet\_wrap()* should not be continuous.

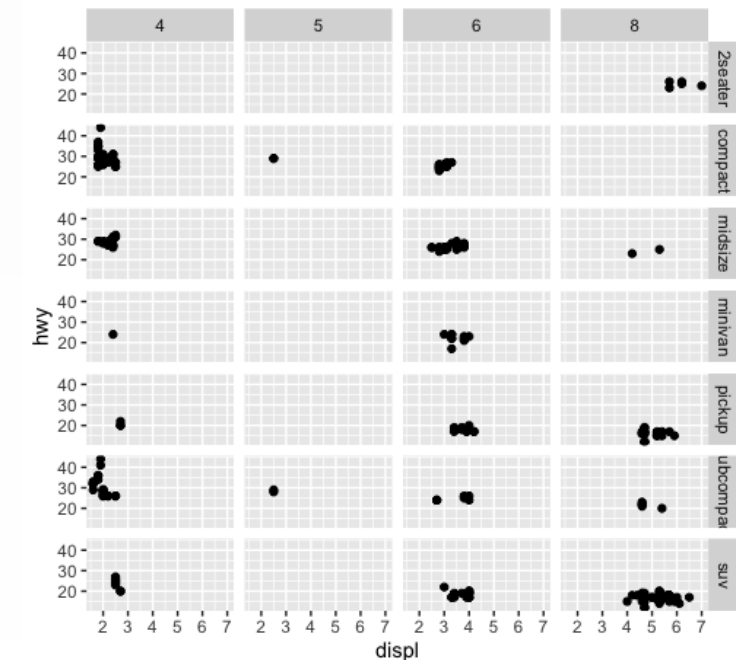


# Facets (cont.)

- You can use `facet_grid()` to facet using two variable.

```
> ggplot(data = mpg) +
+ geom_point(mapping = aes(x = displ, y = hwy)) +
+ facet_grid(class ~ cyl)
```

- The first argument is also a formula, except with two variables.
- Unlike `facet_wrap()`, `facet_grid()` requires two variables separated by .
  - Format: rows variable column variable.



# Geoms

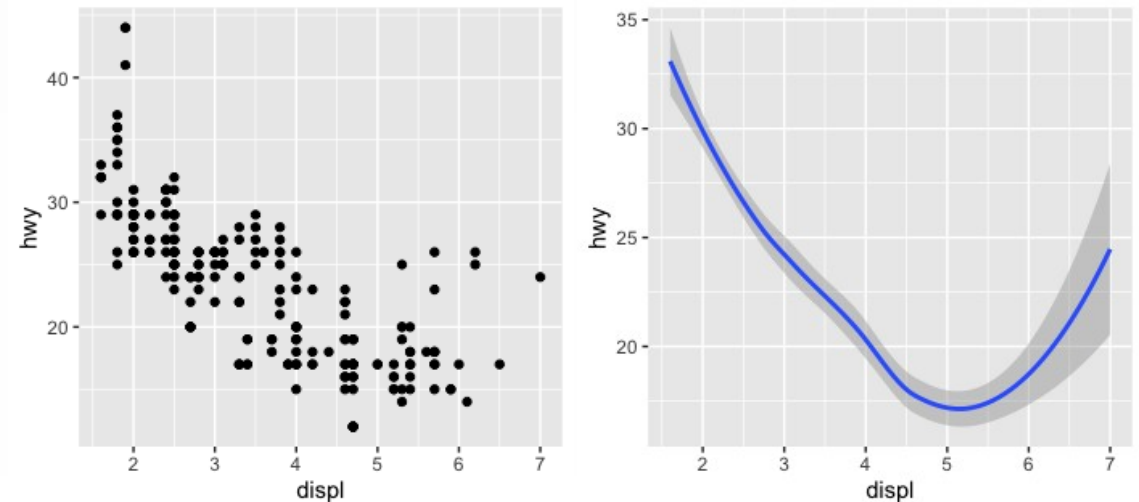
- Geom is a geometrical object that the plot uses to represent the data.
- A scatter plot uses a point geom. Typically, the names are associated –
  - Bar chart uses a bar geom
  - Line chart uses a line geom
  - Boxplots use a boxplot geom
  - There are many others...

This scatterplot can be converted to a smoothed line plot by using `geom_smooth()`.



Not every `aes()` works with every *geom* - for e.g., you can set the shape of a point but not a line.

```
> ggplot(data = mpg) +
+ geom_smooth(mapping = aes(x = displ, y = hwy))
'geom_smooth()' using method = 'loess' and formula 'y ~ x'
```



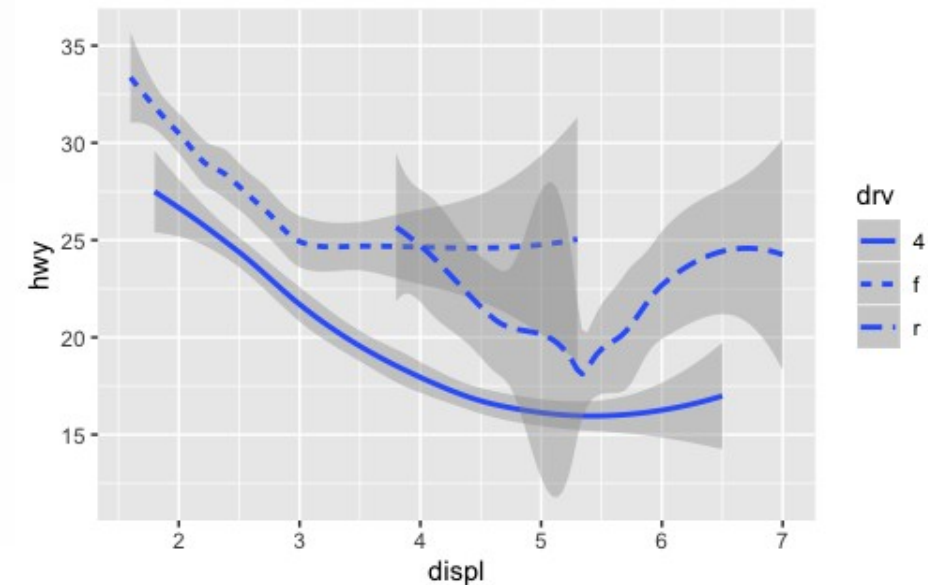
# geom\_smooth()

- The *linetype* argument within *geom\_smooth()* plots a different line with a different linetype for every level of the variable used.
- For e.g., if we want to separate lines by 4wd, front-, or rear-wheel drive.

```
> ggplot(data = mpg) +
+ geom_smooth(mapping = aes(x = displ, y = hwy, linetype = drv))
```

- There are over 40 geoms built-in to ggplot2.
- There are additional geoms available by external packages – <https://www.ggplot2-exts.org>
- Best way to get help

*R-Studio > Help > Cheatsheets > Data Visualization with ggplot2.*





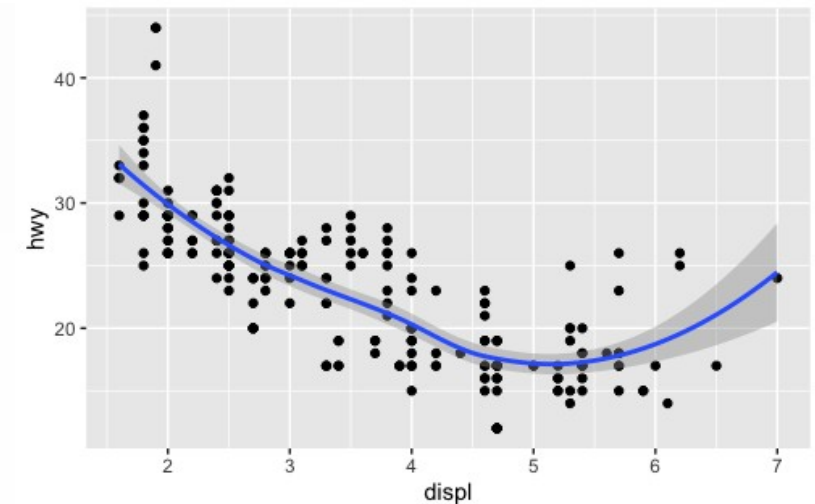
# Multiple *geoms*

- You can add multiple geoms to the same plot. This is handy when you want to see the actual observation and the trend line over it.

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

- Do you see any issues with the above code – code duplication!
- We can place the `aes()` mapping outside the *geom* as global aesthetics.

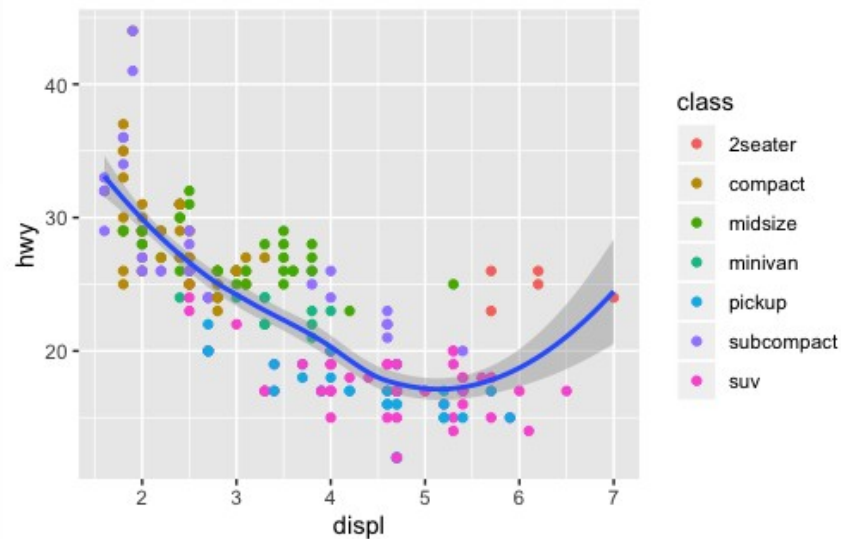
```
> ggplot(data = mpg, mapping = aes(x = displ, y = hwy)) +
+ geom_point() +
+ geom_smooth()
```



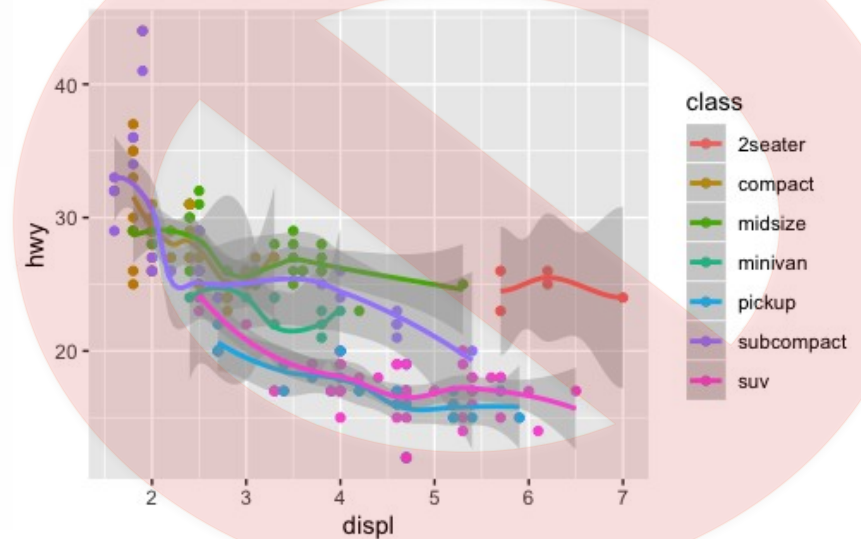
# Multiple *geoms* (cont.)

- You can also override the global aesthetic mappings by placing local aesthetic mappings inside *geoms*. That mapping will work on that layer only.
- For e.g. if we want to color the scatter plot by class, but not the lineplot.

```
> ggplot(data = mpg, mapping = aes(x = displ, y = hwy)) +
+ geom_point(mapping = aes(color = class)) +
+ geom_smooth()
```



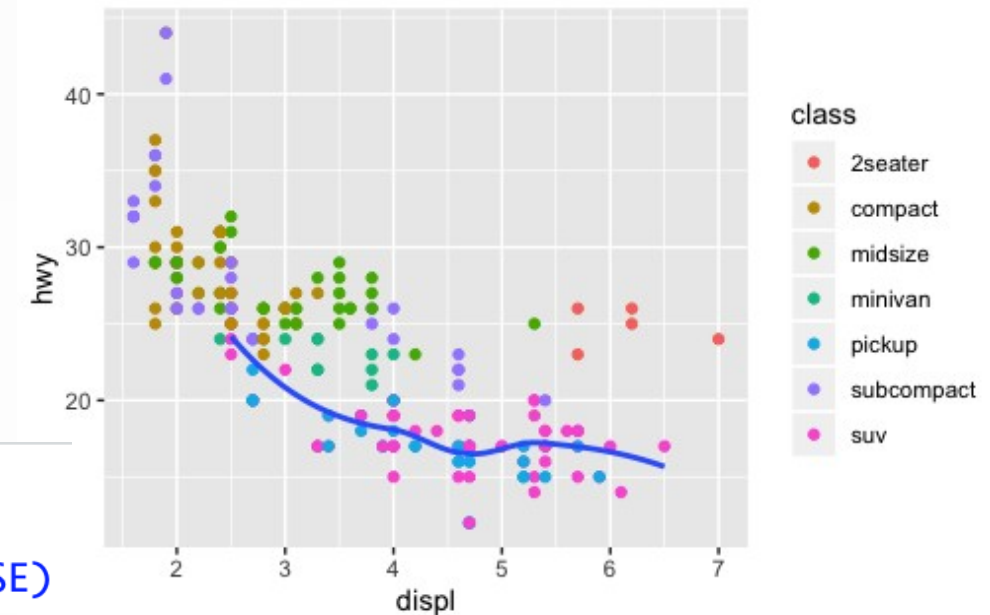
```
> ggplot(data = mpg,
+ mapping = aes(x = displ, y = hwy, color = class)) +
+ geom_point() +
+ geom_smooth()
```



# Multiple *geoms* (cont..)

- You can also specify different data sets for different *geoms*.
- For e.g., if you wanted to draw the smoothed line for only SUVs but wanted to see the data points for the entire data.

```
> ggplot(data = mpg, mapping = aes(x = displ, y = hwy)) +
+ geom_point(mapping = aes(color = class)) +
+ geom_smooth(data = filter(mpg, class == "suv"), se = FALSE)
```

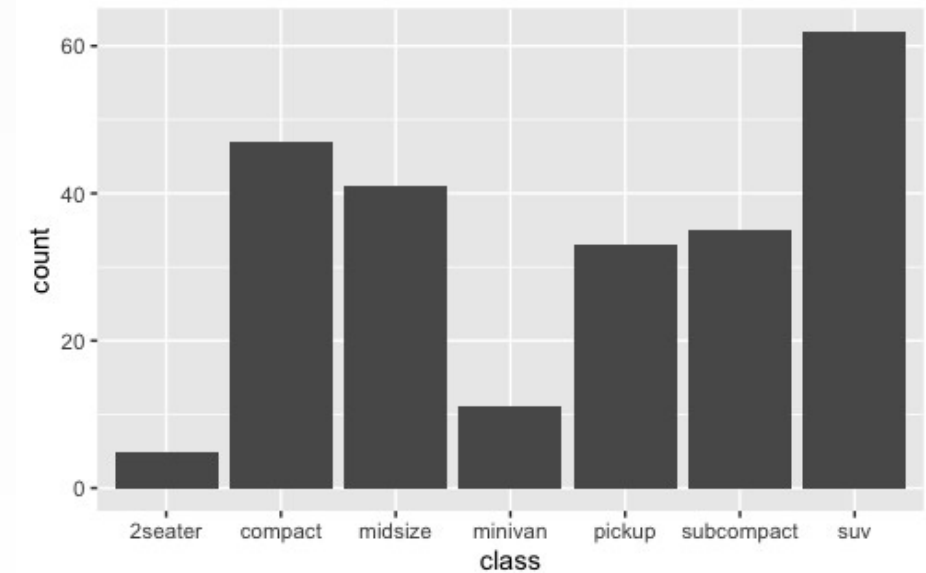


- The local data argument overrides the global mpg data.
  - `filter()` extracts the suv class.
  - `se = FALSE` argument removes the standard error shading.

# *geom\_bar()*

- Bar charts are commonly used for representing a distribution of data – *geom\_bar()*.
- We want to see the distribution of cars across class.
- Notice that the y-axis is count – this is a new computed value from the raw data.
- *geom\_bar()* counts the frequency of each class and uses this on the y-axis.
  - Bar charts, histograms, and frequency plots computes counts and plots bins of counts.
  - Smoothing plots fit a model and then plots fitted values.
  - Boxplots computes summaries and plots the box.

```
ggplot(data = mpg) +
 geom_bar(mapping = aes(x = class))
```



- The default transformations for each of these *geoms* can be found in help under argument *stat*
- For e.g. see `?geom_bar`, `geom_abline`, or `?geom_boxplot`

# Default *geom* and *stat*

Each geometric object can also be made using its stat counterpart.

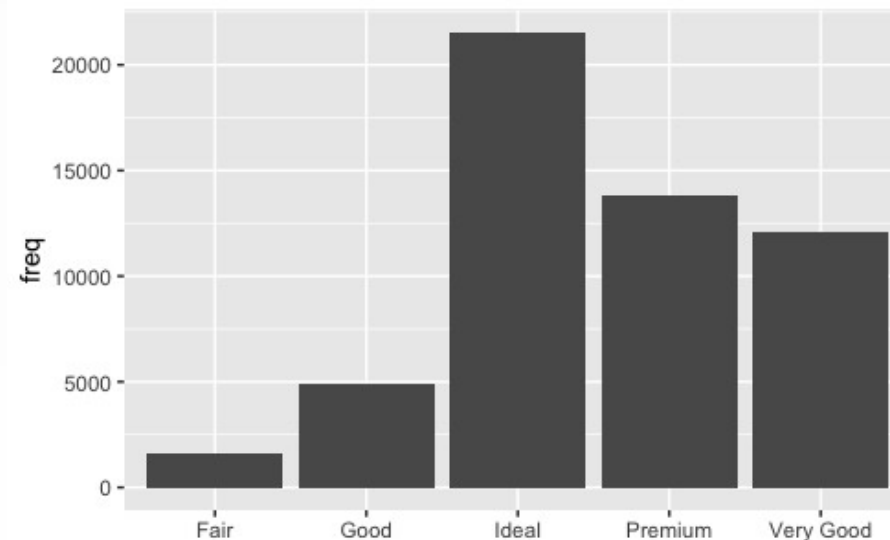
| Geom           | Description                                                      | Default Stat    |
|----------------|------------------------------------------------------------------|-----------------|
| geom_bar()     | Bar chart                                                        | stat_bin()      |
| geom_point()   | Scatterplot                                                      | stat_identity() |
| geom_line()    | Line diagram, connecting observations in order by <b>x-value</b> | stat_identity() |
| geom_boxplot   | Box-and-whisker plot                                             | stat_boxplot()  |
| geom_path      | Line diagram, connecting observations in original order          | stat_identity() |
| geom_smooth    | Add a smoothed conditioned mean                                  | stat_smooth()   |
| geom_histogram | An alias for geom_bar() and stat_bin()                           | stat_bin()      |

# When to Override Default *stat*

- Suppose we wanted to show the proportions on the y-axis.
- For e.g., create a data frame about diamond cuts called *demo*.

```
> demo <- tribble(
+ ~cut, ~freq,
+ "Fair", 1610,
+ "Good", 4906,
+ "Very Good", 12082,
+ "Premium", 13791,
+ "Ideal", 21551
+)
```

```
> ggplot(data = demo) +
+ geom_bar(mapping = aes(x = cut, y = freq), stat = "identity")
```



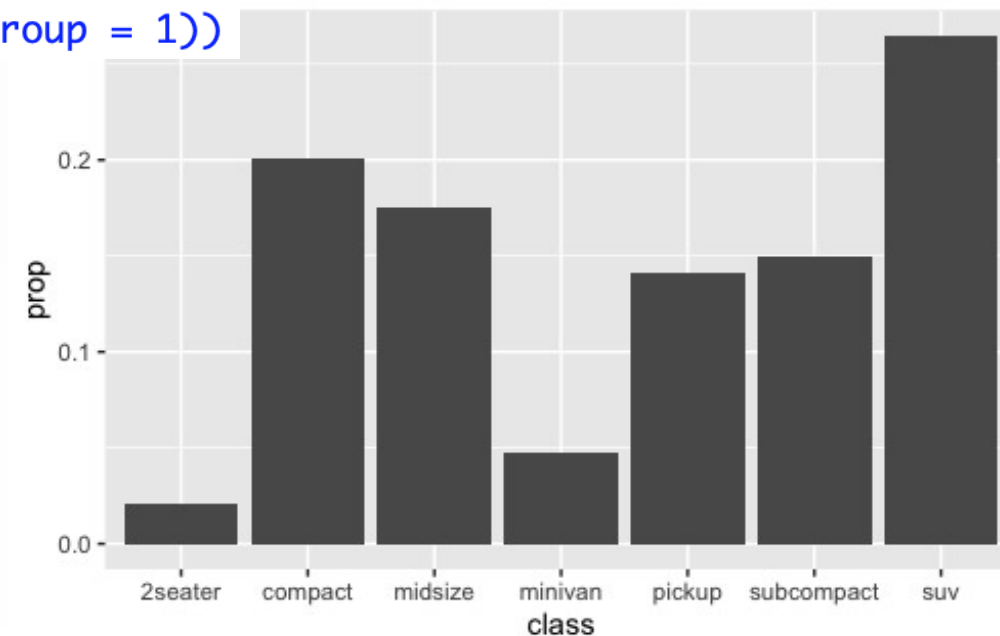
- Using `stat = "identity"` allows me to plot the summarized values directly.

# When to Override Default *stat* (cont.)

- Now on the y-axis, if we want to show proportions instead of frequency.
- We can use the *mpg* data – based on class.

```
> ggplot(data = mpg) +
+ geom_bar(mapping = aes(x = class, y = stat(prop), group = 1))
```

- $y = \text{stat(prop)}$  plots the proportion on y-axis.
- We tell ggplot to compute proportion by taking all the classes into a single group, using  $\text{group} = 1$ .
  - If you take that out, all classes will have  $\text{prop} = 1$ .

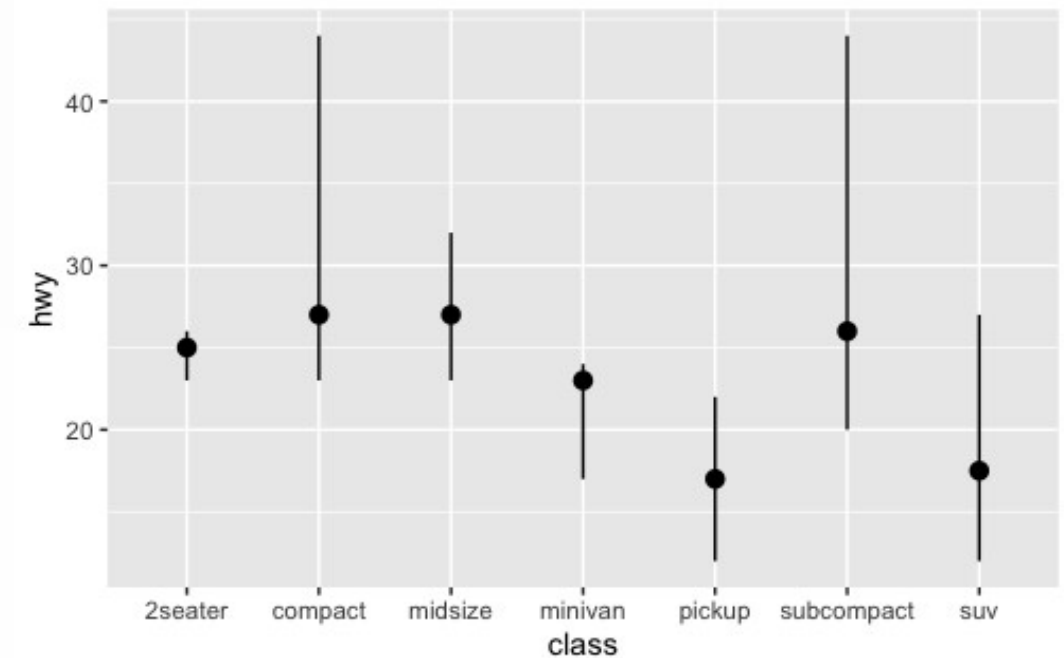


# When to Override Default *stat* (cont..)

- Or else, maybe we want to plot summary statistics from the data.
  - Use `stat_summary()` – summarizes *y* values for each *x* value.

```
> ggplot(data = mpg) +
+ stat_summary(
+ mapping = aes(x = class, y = hwy),
+ fun.ymin = min,
+ fun.ymax = max,
+ fun.y = median
+)
```

- `ggplot()` provides over 20 *stat* functions. You can get help for each using `?`, for e.g. `?stat_bin`.

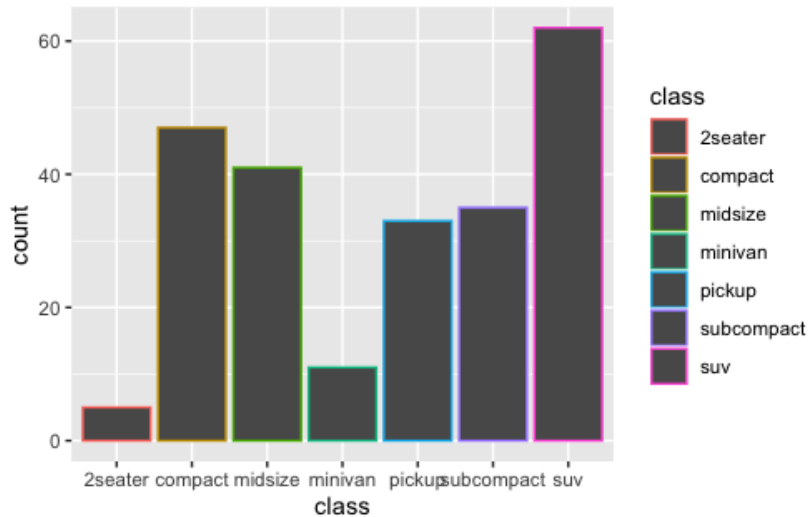




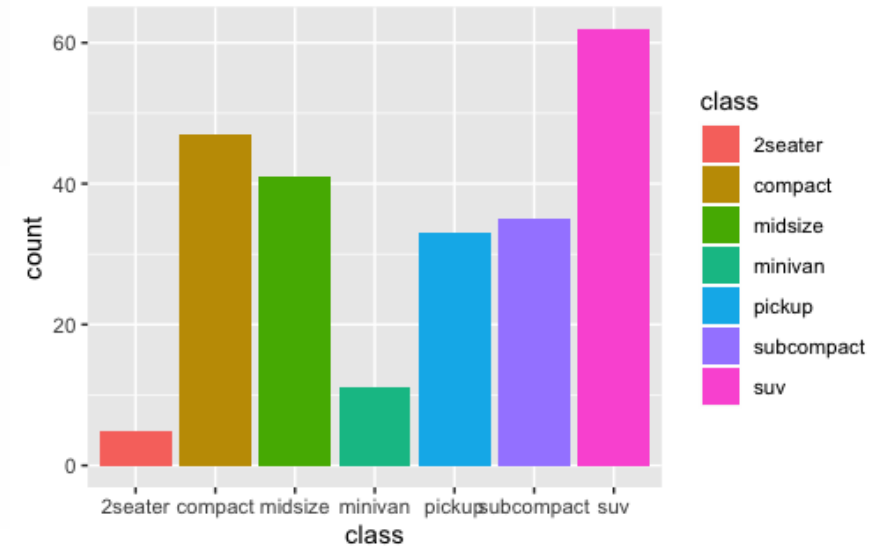
# *geom\_bar() (cont.)*

You can also color bar charts using color or fill arguments.

```
> ggplot(data = mpg) +
+ geom_bar(mapping = aes(x = class, colour = class))
```



```
> ggplot(data = mpg) +
+ geom_bar(mapping = aes(x = class, fill = class))
```

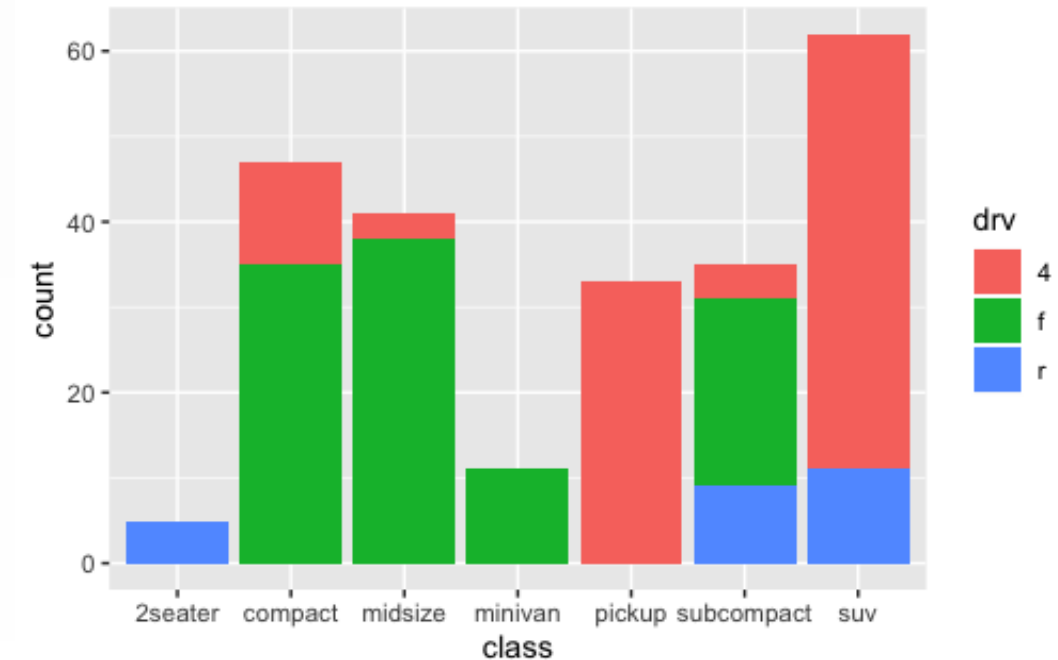
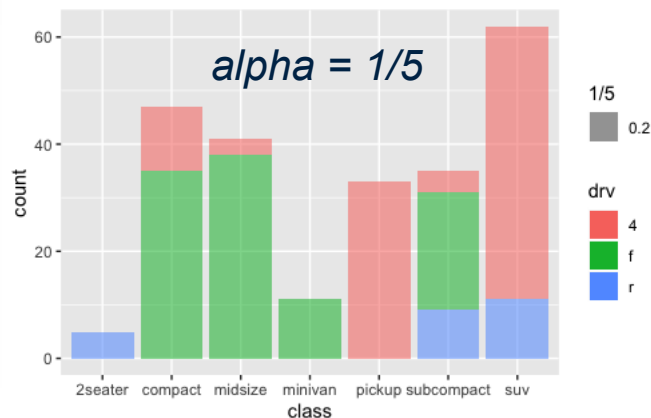


## *geom\_bar()* (cont..)

- If you change the fill variable to something other than the x variable, you can see the distribution within each bar.
- For e.g. see the distribution of *drv* within each *class*

```
> ggplot(data = mpg) +
+ geom_bar(mapping = aes(x = class, fill = drv))
```

- You can also control the transparency of the bars using *alpha* inside *aes()*.

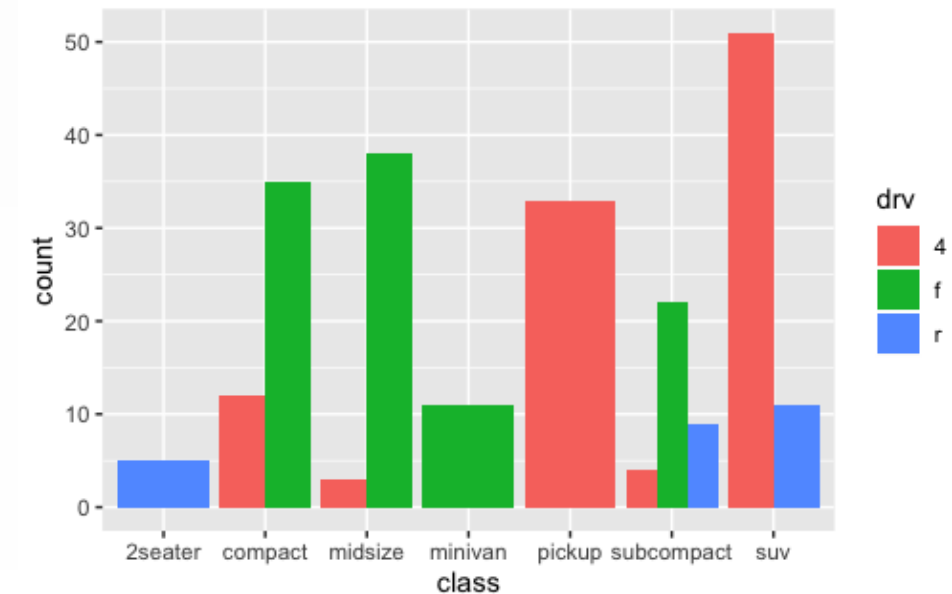


# Unstacked Bar Chart

- If you don't want stacked bar chart, but rather next to each other.
- Use position = "dodge"

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

- Note that the position argument is outside `aes()` but inside `geom_bar()`.



# *geom\_histogram()*

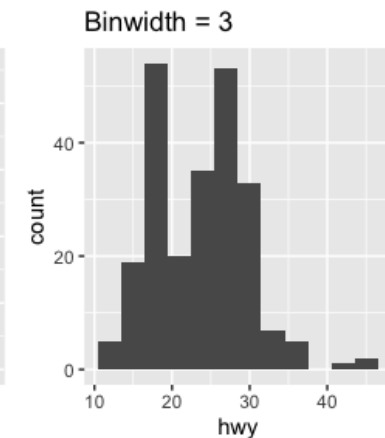
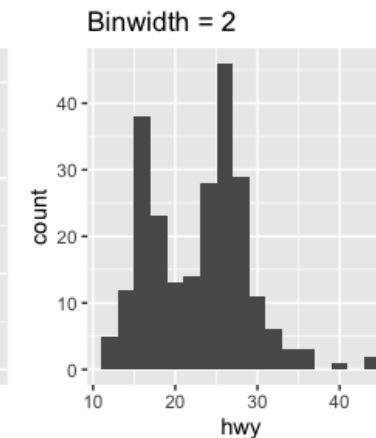
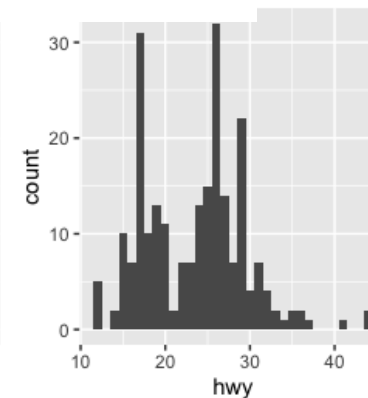
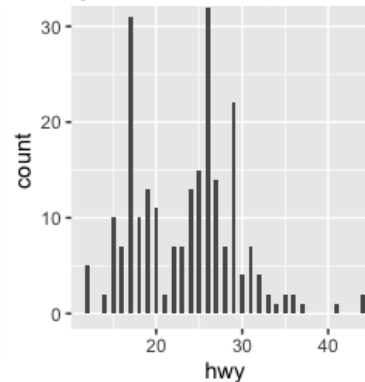
- Another popular type of plot used in statistical analysis are histograms. They show the distribution of the data.
- In *geom\_histogram()*, you can specify the *binwidth* to get a better understanding of the distribution of the data.

```
> ggplot(data=mpg) +
+ geom_histogram(binwidth = 2, mapping = aes(x = hwy))
```

- Do *install.packages("gridExtra")* and load it - makes multiple plots on a grid easily.
- You can assign plots to objects and plot multiple objects on a grid.

```
> plot1 <- ggplot(data=mpg) +
+ geom_histogram(binwidth = 0.5, mapping = aes(x = hwy)) +
+ labs(title = "Binwidth = 0.5")
```

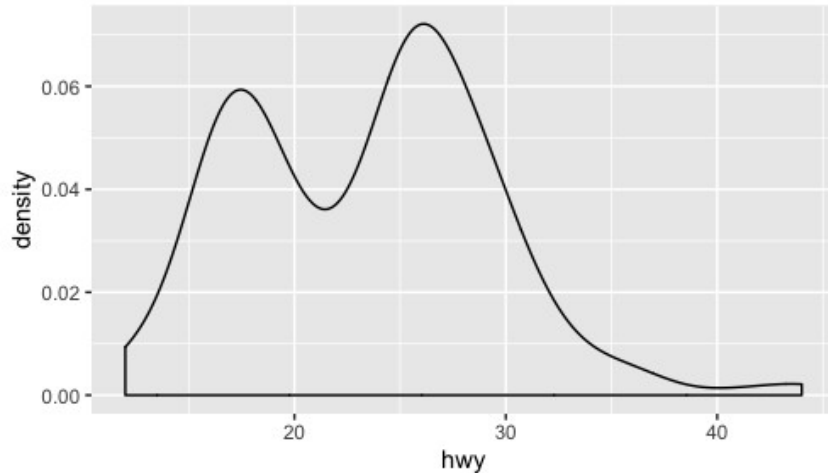
```
> plot1 <- ggplot(data=mpg) +
+ geom_histogram(binwidth = 0.5, mapping = aes(x = hwy)) +
+ labs(title = "Binwidth = 0.5")
> plot2 <- ggplot(data=mpg) +
+ geom_histogram(binwidth = 1, mapping = aes(x = hwy)) +
+ labs(title = "Binwidth = 1")
> plot3 <- ggplot(data=mpg) +
+ geom_histogram(binwidth = 2, mapping = aes(x = hwy)) +
+ labs(title = "Binwidth = 2")
> plot4 <- ggplot(data=mpg) +
+ geom_histogram(binwidth = 3, mapping = aes(x = hwy)) +
+ labs(title = "Binwidth = 3")
> grid.arrange(plot1, plot2, plot3, plot4, ncol=4)
```



# geom\_density() and geom\_segment()

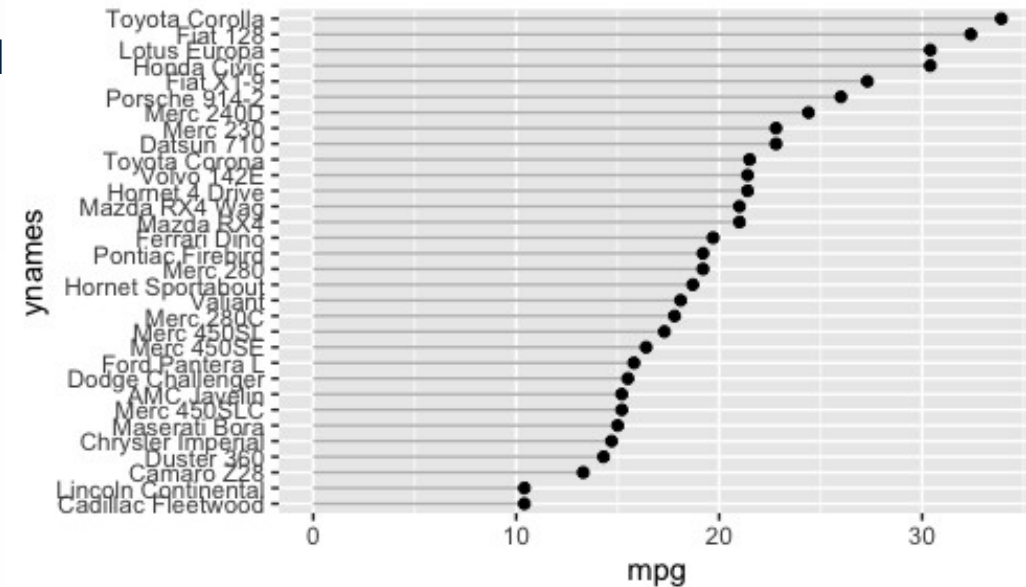
- A smoothed histogram can be plotted using a density plot.

```
> ggplot(data=mpg) +
+ geom_density(mapping = aes(x = hwy))
```



```
> mtcars %>%
+ mutate(carnames = rownames(mtcars)) %>%
+ arrange(mpg) %>%
+ mutate(ynames = factor(carnames, levels = carnames)) %>%
+ ggplot(aes(x = mpg, y = ynames)) +
+ geom_segment(aes(x = 0, y = ynames, xend = mpg, yend = ynames), color = "grey") +
+ geom_point()
```

- Sometimes, we want to see a hybrid between bar chart and scatterplot.
- Plot *mtcars* data sorted based on mpg, and then all the cars plotted along with their values.

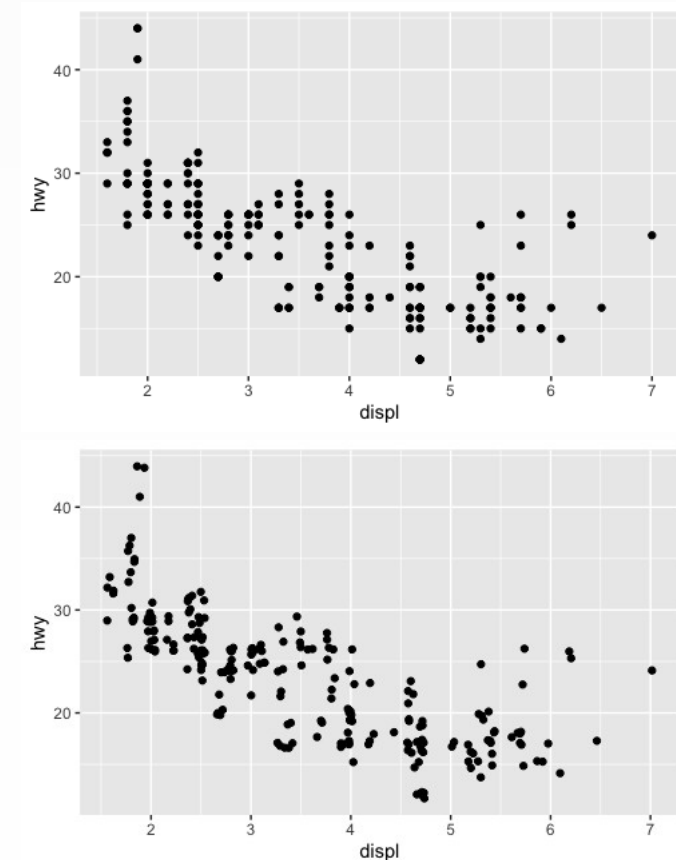


# Jittered Scatterplot

- Often, observations overlap each other on the plot.
  - For e.g., see the first scatterplot we made.
  - This plot shows 126 points even though there are 234 observations.
  - A lot of overlap: problem known as overplotting!
- The argument *position* = "jitter" overcomes this by slightly moving overlapping observations.

```
> ggplot(data = mpg) +
+ geom_point(mapping = aes(x = displ, y = hwy),
+ position = "jitter")
```

- Less accurate in small scale but reveals better insights about the data in the big picture.



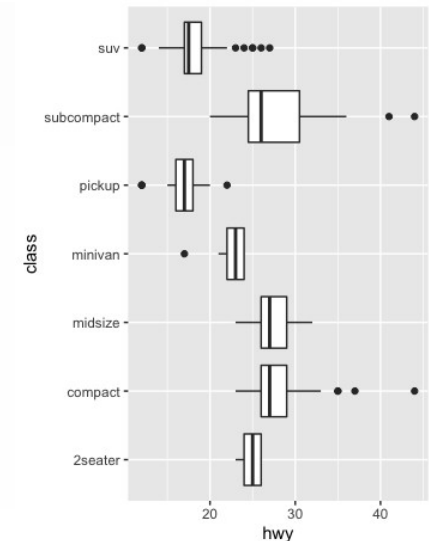
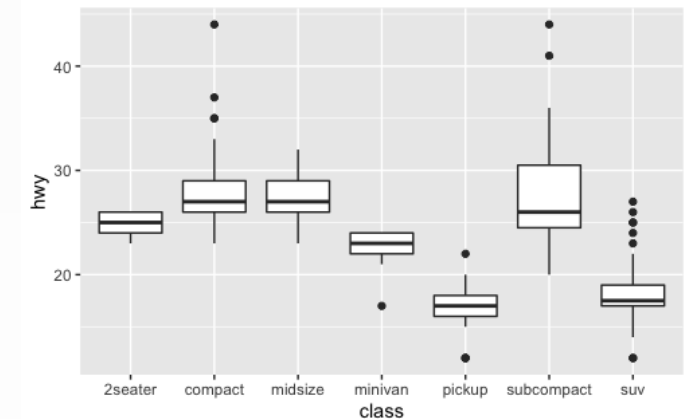
# Coordinate System

- By default, Cartesian coordinates.
- `coord_flip()` switches the x and y axes.
- Suppose we made a boxplot of highway mileage across the classes.

```
> ggplot(data = mpg, mapping = aes(x = class, y = hwy)) +
+ geom_boxplot()
```

- Then, we wanted to switch this into a horizontal boxplot.

```
> ggplot(data = mpg, mapping = aes(x = class, y = hwy)) +
+ geom_boxplot() +
+ coord_flip()
```

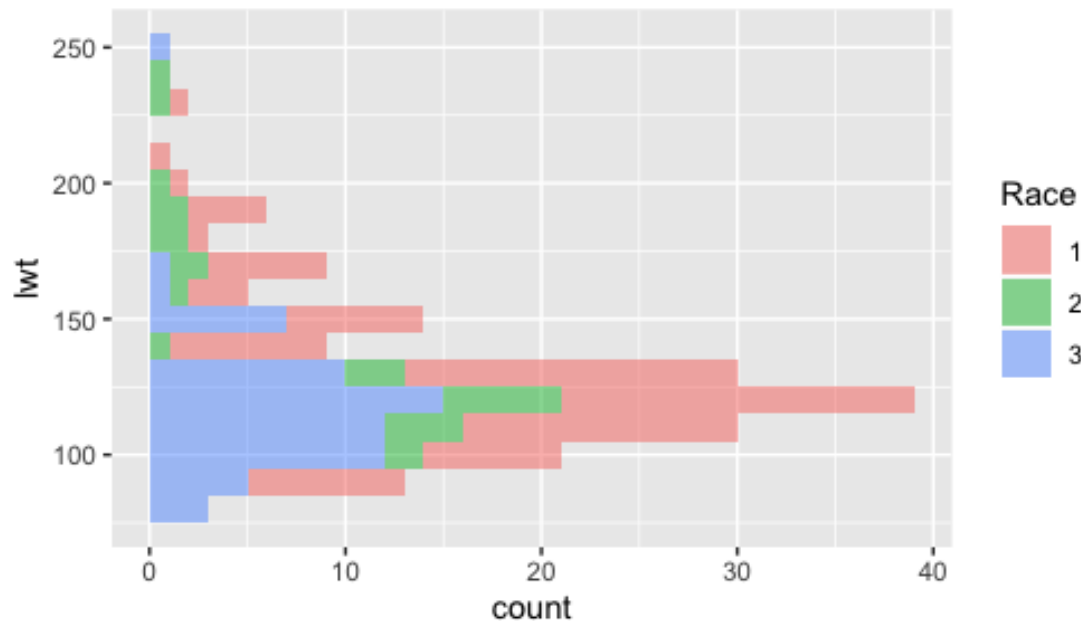




# Examples

Install the *MASS* package and load the *"birthwt"* data. Do *?birthwt* for more details.

1. Make the plot shown below.
2. Transparency is set of 50%, and histogram uses binwidth of 10.



Hints: to change legend title: `scale_fill_discrete(name="Race")`

*If you're getting gradient coloring, think about the class of data.*

# Examples (cont.)

- Reproduce the plot below. Hint – remember *facet\_wrap()* and use *binwidth = 10*!
- Again, make sure the class of variables are correct.

