Selected Exercises for STT 2860

Material from R for Data Science (2e)

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Preface

This document takes selected Exercises from R for Data Science (2e) and puts them in a Quarto book so you can modify the files to add your answers to the questions.

To learn more about Quarto books please visit https://quarto.org/docs/books.

1 Exercises (Chapter 1)

1. How many rows are in penguins? How many columns?

```
Answer

library(tidyverse)
library(palmerpenguins)
# Your R Code here

Your answer here.
```

2. What does the bill_depth_mm variable in the penguins data frame describe? Read the help for ?penguins to find out.

Answer

Type your answer here.

3. Make a scatterplot of bill_depth_mm vs. bill_length_mm. That is, make a scatterplot with bill_depth_mm on the y-axis and bill_length_mm on the x-axis. Describe the relationship between these two variables.

```
Answer

# Your R code here

Your answer here.
```

4. What happens if you make a scatterplot of species vs. bill_depth_mm? What might be a better choice of geom?

```
Answer

# Your R code here

Your answer here.
```

```
# Your R code here
```

5. Why does the following give an error and how would you fix it?

```
library(tidyverse)
ggplot(data = penguins) +
  geom_point()
```

Answer

Your answer here.

Correct code here

6. What does the na.rm argument do in geom_point()? What is the default value of the argument? Create a scatterplot where you successfully use this argument set to TRUE.

Answer

Your answer here.

Your R code here

7. Add the following caption to the plot you made in the previous exercise: "Data come from the palmerpenguins package." Hint: Take a look at the documentation for labs().

Answer # Your R code here

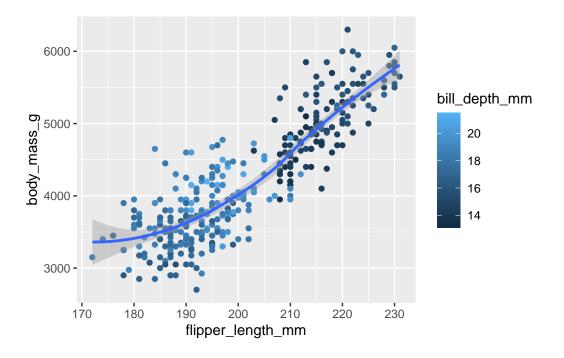
8. Recreate the following visualization. What aesthetic should bill_depth_mm be mapped to? And should it be mapped at the global level or at the geom level?

```
Answer

# Your R Code here

Your answer here.
```

9. Run this code in your head and predict what the output will look like. Then, run the code in R and check your predictions.



```
ggplot(
  data = penguins,
  mapping = aes(x = flipper_length_mm, y = body_mass_g, color = island)
) +
    geom_point() +
    geom_smooth(se = FALSE)
```

```
Answer

Your answer here.

# Your R code here
```

10. Will these two graphs look different? Why/why not?

```
ggplot(
  data = penguins,
  mapping = aes(x = flipper_length_mm, y = body_mass_g)
) +
  geom_point() +
  geom_smooth()
```

```
ggplot() +
  geom_point(
    data = penguins,
    mapping = aes(x = flipper_length_mm, y = body_mass_g)
) +
  geom_smooth(
    data = penguins,
    mapping = aes(x = flipper_length_mm, y = body_mass_g)
)
```

```
Answer

Your answer here.

library(patchwork)
# Your R code here
```

11. Make a bar plot of species of penguins, where you assign species to the y aesthetic. How is this plot different?

```
Answer

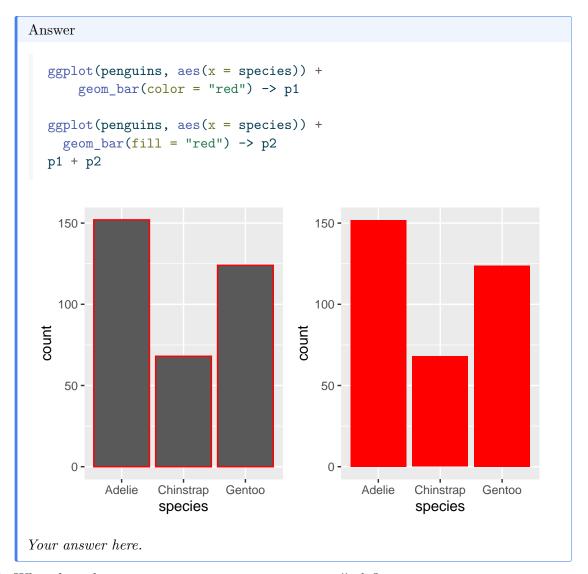
Your answer here.

# Your R code here
```

12. How are the following two plots different? Which aesthetic, color or fill, is more useful for changing the color of bars?

```
ggplot(penguins, aes(x = species)) +
  geom_bar(color = "red")

ggplot(penguins, aes(x = species)) +
  geom_bar(fill = "red")
```



13. What does the bins argument in geom_histogram() do?

Answer

Your answer here.

14. Make a histogram of the carat variable in the diamonds dataset that is available when you load the tidyverse package. Experiment with different binwidths. What binwidth reveals the most interesting patterns?

Answer

Your R code here

Your answer here.

15. The mpg data frame that is bundled with the ggplot2 package contains 234 observations collected by the US Environmental Protection Agency on 38 car models. Which variables in mpg are categorical? Which variables are numerical? (Hint: Type ?mpg to read the documentation for the dataset.) How can you see this information when you run mpg?

Answer

Your R code here

Your answer here.

16. Make a scatterplot of hwy vs. displ using the mpg data frame. Next, map a third, numerical variable to color, then size, then both color and size, then shape. How do these aesthetics behave differently for categorical vs. numerical variables?

Answer

Your R code here

Your answer here.

17. In the scatterplot of hwy vs. displ, what happens if you map a third variable to linewidth?

Answer

Your R code here

Your answer here.

18. What happens if you map the same variable to multiple aesthetics?

Answer

Your R code here

Your answer here.

19. Make a scatterplot of bill_depth_mm vs. bill_length_mm and color the points by species. What does adding coloring by species reveal about the relationship between these two variables? What about faceting by species?

```
Answer

# Your R code here

Your answer here.
```

20. Why does the following yield two separate legends? How would you fix it to combine the two legends?

```
ggplot(
  data = penguins,
  mapping = aes(
    x = bill_length_mm, y = bill_depth_mm,
    color = species, shape = species
)
) +
  geom_point() +
  labs(color = "Species")
```

```
Answer

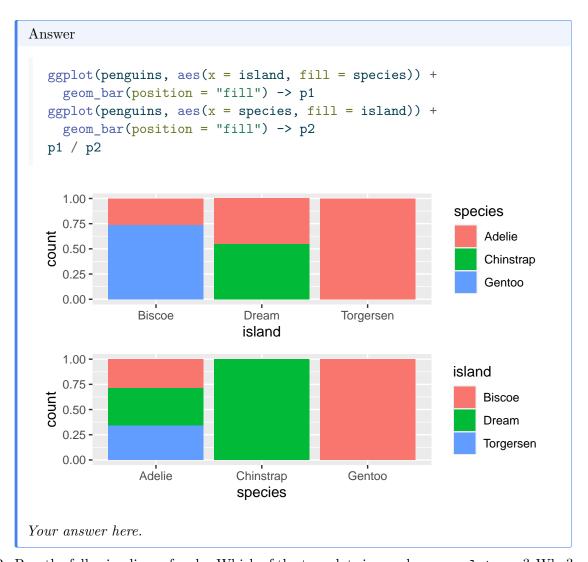
# Your R fix here

Your answer here.
```

21. Create the two following stacked bar plots. Which question can you answer with the first one? Which question can you answer with the second one?

```
ggplot(penguins, aes(x = island, fill = species)) +
  geom_bar(position = "fill")

ggplot(penguins, aes(x = species, fill = island)) +
  geom_bar(position = "fill")
```



22. Run the following lines of code. Which of the two plots is saved as mpg-plot.png? Why?

```
ggplot(mpg, aes(x = class)) +
    geom_bar()
ggplot(mpg, aes(x = cty, y = hwy)) +
    geom_point()
ggsave("mpg-plot.png")
```

```
Answer

# Your R code here
```

Your answer here.

23. What do you need to change in the code above to save the plot as a PDF instead of a PNG? How could you find out what types of image files would work in ggsave()?

Answer

Your answer here.

2 Exercises (Chapter 2)

1. Why does this code not work?

```
my_variable <- 10
my_variable</pre>
```

Look carefully! (This may seem like an exercise in pointlessness, but training your brain to notice even the tiniest difference will pay off when programming.)

Answer

Your answer here.

2. Tweak each of the following R commands so that they run correctly:

```
libary(todyverse)
ggplot(dTA = mpg) +
  geom_point(maping = aes(x = displ y = hwy)) +
  geom_smooth(method = "lm)
```

Answer

Your R code here

3. Press Option + Shift + K / Alt + Shift + K. What happens? How can you get to the same place using the menus?

Answer

 $Your\ answer\ here.$

4. Let's revisit an exercise from "Saving Your Plots". Run the following lines of code. Which of the two plots is saved as mpg-plot.png? Why?

```
my_bar_plot <- ggplot(mpg, aes(x = class)) +
    geom_bar()
my_scatter_plot <- ggplot(mpg, aes(x = cty, y = hwy)) +
    geom_point()
ggsave(filename = "mpg-plot.png", plot = my_bar_plot)</pre>
```

```
Answer

my_bar_plot <- ggplot(mpg, aes(x = class)) +
    geom_bar()
my_scatter_plot <- ggplot(mpg, aes(x = cty, y = hwy)) +
    geom_point()
ggsave(filename = "mpg-plot.png", plot = my_bar_plot)

Saving 5.5 x 3.5 in image
Your answer here.</pre>
```

3 Exercises (Chapter 3)

- 1. In a single pipeline for each condition, find all flights that meet the condition:
 - Had an arrival delay of two or more hours

```
Answer

library(tidyverse)
library(nycflights13)
# Your code here
```

• Flew to Houston (IAH or HOU)

```
Answer

# Your code here
```

• Were operated by United, American, or Delta

```
Answer

# Your code here
```

• Departed in summer (July, August, and September)

```
Answer

# Your code here
```

• Arrived more than two hours late, but didn't leave late.

Answer # Your code here

• Were delayed by at least an hour, but made up over 30 minutes in flight

Answer

Your code here

2. Sort flights to find the flights with longest departure delays. Find the flights that left earliest in the morning.

Answer

Your code here

3. Sort flights to find the fastest flights. (Hint: Try including a math calculation inside of your function.)

Answer

Your code here

4. Was there a flight on every day of 2013?

Answer

Your code here

Your text answer here.

5. Which flights traveled the farthest distance? Which traveled the least distance?

Answer

Your code here

6. Does it matter what order you used filter() and arrange() if you're using both? Why/why not? Think about the results and how much work the functions would have

to do.

Answer

Your text answer here.

7. Compare dep_time, sched_dep_time, and dep_delay. How would you expect those three numbers to be related?

Answer # Your code here Your text answer here.

8. Brainstorm as many ways as possible to select dep_time, dep_delay, arr_time, and arr_delay from flights.

```
Answer
# Your code here
```

9. What happens if you specify the name of the same variable multiple times in a select() call?

```
Answer

# Your code here

Your text answer here.
```

10. What does the any_of() function do? Why might it be helpful in conjunction with this vector?

```
variables <- c("year", "month", "day", "dep_delay", "arr_delay")
# Try below first
flights |>
   select(variables)
```

Warning: Using an external vector in selections was deprecated in tidyselect 1.1.0.
i Please use `all_of()` or `any_of()` instead.
 # Was:
 data %>% select(variables)

Now: data %>% select(all_of(variables))

See https://tidyselect.r-lib.org/reference/faq-external-vector.html.

A tibble: 336,776 x 5

	year	${\tt month}$	day	dep_delay	arr_delay
	<int></int>	<int></int>	<int></int>	<dbl></dbl>	<dbl></dbl>
1	2013	1	1	2	11
2	2013	1	1	4	20
3	2013	1	1	2	33
4	2013	1	1	-1	-18
5	2013	1	1	-6	-25
6	2013	1	1	-4	12
7	2013	1	1	-5	19
8	2013	1	1	-3	-14
9	2013	1	1	-3	-8
10	2013	1	1	-2	8

i 336,766 more rows

```
# Or
flights |>
select(any_of(variables))
```

A tibble: 336,776 x 5

	year	month	day	<pre>dep_delay</pre>	arr_delay
	<int></int>	<int></int>	<int></int>	<dbl></dbl>	<dbl></dbl>
1	2013	1	1	2	11
2	2013	1	1	4	20
3	2013	1	1	2	33
4	2013	1	1	-1	-18
5	2013	1	1	-6	-25
6	2013	1	1	-4	12
7	2013	1	1	-5	19
8	2013	1	1	-3	-14
9	2013	1	1	-3	-8
10	2013	1	1	-2	8

i 336,766 more rows

Answer

Your text answer here.

11. Does the result of running the following code surprise you? How do the select helpers deal with upper and lower case by default? How can you change that default?

```
flights |>
select(contains("TIME"))
```

```
Answer
  flights |>
     select(contains("TIME"))
# A tibble: 336,776 x 6
   dep_time sched_dep_time arr_time sched_arr_time air_time time_hour
      <int>
                      <int>
                                <int>
                                                <int>
                                                          <dbl> <dttm>
 1
        517
                        515
                                  830
                                                  819
                                                            227 2013-01-01 05:00:00
 2
        533
                        529
                                  850
                                                  830
                                                            227 2013-01-01 05:00:00
 3
        542
                        540
                                  923
                                                  850
                                                            160 2013-01-01 05:00:00
 4
        544
                        545
                                 1004
                                                 1022
                                                            183 2013-01-01 05:00:00
 5
        554
                        600
                                  812
                                                  837
                                                            116 2013-01-01 06:00:00
 6
        554
                        558
                                  740
                                                  728
                                                            150 2013-01-01 05:00:00
 7
                                                            158 2013-01-01 06:00:00
        555
                        600
                                  913
                                                  854
 8
                                  709
                                                  723
                                                             53 2013-01-01 06:00:00
        557
                        600
 9
        557
                                  838
                                                            140 2013-01-01 06:00:00
                        600
                                                  846
                                                            138 2013-01-01 06:00:00
10
        558
                        600
                                  753
                                                  745
# i 336,766 more rows
Your text answer here.
```

12. Rename air_time to air_time_min to indicate units of measurement and move it to the beginning of the data frame.

```
Answer

# Your code here
```

13. Why doesn't the following work, and what does the error mean?

```
flights |>
  select(tailnum) |>
  arrange(arr_delay)
```

Error in `arrange()`:

```
i In argument: `..1 = arr_delay`.
Caused by error:
! object 'arr_delay' not found
  flights |>
    select(tailnum)
# A tibble: 336,776 x 1
  tailnum
   <chr>>
1 N14228
2 N24211
3 N619AA
4 N804JB
5 N668DN
6 N39463
7 N516JB
8 N829AS
9 N593JB
10 N3ALAA
# i 336,766 more rows
```

Answer

Your text answer here.

14. Which carrier has the worst average delays? Challenge: can you disentangle the effects of bad airports vs. bad carriers? Why/why not? (Hint: think about flights |> group_by(carrier, dest) |> summarize(n()))

Answer

Your code here

Your text answer here.

15. Find the flights that are most delayed upon departure from each destination.

Answer

Your code here

Your text answer here.

16. How do delays vary over the course of the day. Illustrate your answer with a plot.

```
# Your code here

Your text answer here.
```

17. What happens if you supply a negative n to slice_min() and friends?

```
Answer
  flights |>
    slice_min(dep_delay, n = -5) |>
    relocate(dep_delay)
# A tibble: 336,776 x 19
   dep_delay year month
                            day dep_time sched_dep_time arr_time sched_arr_time
       <dbl> <int> <int> <int>
                                   <int>
                                                   <int>
                                                             <int>
                                                                            <int>
         -43 2013
                       12
                              7
                                    2040
                                                    2123
                                                                40
                                                                             2352
 2
         -33 2013
                        2
                              3
                                    2022
                                                    2055
                                                              2240
                                                                             2338
 3
         -32 2013
                       11
                             10
                                    1408
                                                    1440
                                                             1549
                                                                             1559
         -30 2013
                             11
                                    1900
                                                    1930
                                                              2233
                                                                             2243
 5
         -27 2013
                             29
                                    1703
                                                    1730
                                                             1947
                                                                             1957
                        1
 6
         -26 2013
                              9
                                     729
                                                     755
                                                             1002
                                                                              955
                        8
 7
         -25 2013
                       10
                             23
                                    1907
                                                    1932
                                                             2143
                                                                             2143
 8
         -25 2013
                        3
                             30
                                    2030
                                                    2055
                                                             2213
                                                                             2250
 9
         -24 2013
                              2
                                                    1455
                        3
                                    1431
                                                              1601
                                                                             1631
10
         -24 2013
                        5
                              5
                                     934
                                                     958
                                                              1225
                                                                             1309
# i 336,766 more rows
# i 11 more variables: arr_delay <dbl>, carrier <chr>, flight <int>,
    tailnum <chr>, origin <chr>, dest <chr>, air_time <dbl>, distance <dbl>,
    hour <dbl>, minute <dbl>, time_hour <dttm>
  flights |>
    slice_min(dep_delay, n = 5) |>
    relocate(dep_delay)
# A tibble: 5 x 19
  dep_delay year month
                           day dep_time sched_dep_time arr_time sched_arr_time
      <dbl> <int> <int> <int>
                                  <int>
                                                  <int>
                                                            <int>
                                                                           <int>
                                   2040
                                                   2123
                                                                            2352
1
        -43
            2013
                      12
                             7
                                                               40
```

```
2
                              3
                                    2022
        -33
              2013
                       2
                                                     2055
                                                              2240
                                                                               2338
3
        -32
              2013
                             10
                                    1408
                                                     1440
                                                              1549
                                                                               1559
                      11
        -30
              2013
                                    1900
                                                     1930
                                                              2233
4
                       1
                             11
                                                                               2243
5
        -27
              2013
                       1
                             29
                                    1703
                                                     1730
                                                              1947
                                                                               1957
# i 11 more variables: arr delay <dbl>, carrier <chr>, flight <int>,
    tailnum <chr>, origin <chr>, dest <chr>, air_time <dbl>, distance <dbl>,
    hour <dbl>, minute <dbl>, time_hour <dttm>
  flights |>
    slice_max(dep_delay, n = -5) |>
    relocate(dep_delay)
# A tibble: 336,776 x 19
   dep delay year month
                             day dep_time sched_dep_time arr_time sched_arr_time
       <dbl> <int> <int> <int>
                                    <int>
                                                     <int>
                                                               <int>
                                                                               <int>
 1
        1301
               2013
                                       641
                                                       900
                                                               1242
                                                                                1530
 2
        1137
               2013
                         6
                              15
                                      1432
                                                      1935
                                                               1607
                                                                                2120
 3
        1126
              2013
                         1
                              10
                                      1121
                                                      1635
                                                               1239
                                                                                1810
 4
        1014
              2013
                        9
                              20
                                      1139
                                                      1845
                                                               1457
                                                                                2210
 5
        1005
              2013
                        7
                              22
                                      845
                                                      1600
                                                               1044
                                                                                1815
 6
         960 2013
                        4
                              10
                                     1100
                                                      1900
                                                               1342
                                                                                2211
 7
         911
              2013
                        3
                              17
                                      2321
                                                       810
                                                                135
                                                                                1020
 8
                              27
         899
              2013
                         6
                                      959
                                                      1900
                                                               1236
                                                                                2226
                        7
 9
         898
              2013
                              22
                                      2257
                                                       759
                                                                121
                                                                                1026
10
         896 2013
                       12
                               5
                                      756
                                                      1700
                                                                                2020
                                                               1058
# i 336,766 more rows
# i 11 more variables: arr delay <dbl>, carrier <chr>, flight <int>,
#
    tailnum <chr>, origin <chr>, dest <chr>, air_time <dbl>, distance <dbl>,
    hour <dbl>, minute <dbl>, time hour <dttm>
  flights |>
     slice_max(dep_delay, n = 5) \mid >
    relocate(dep_delay)
# A tibble: 5 x 19
  dep_delay year month
                            day dep_time sched_dep_time arr_time sched_arr_time
                                                             <int>
      <dbl> <int> <int> <int>
                                   <int>
                                                    <int>
                                                                              <int>
1
       1301
              2013
                       1
                              9
                                      641
                                                      900
                                                              1242
                                                                               1530
2
       1137
              2013
                       6
                             15
                                    1432
                                                     1935
                                                              1607
                                                                               2120
3
       1126
              2013
                             10
                                                     1635
                                                                               1810
                       1
                                    1121
                                                              1239
4
       1014
              2013
                       9
                             20
                                    1139
                                                     1845
                                                              1457
                                                                               2210
```

```
5 1005 2013 7 22 845 1600 1044 1815 # i 11 more variables: arr_delay <dbl>, carrier <chr>, flight <int>, # tailnum <chr>, origin <chr>, dest <chr>, air_time <dbl>, distance <dbl>, # hour <dbl>, minute <dbl>, time_hour <dttm>

Your text answer here.
```

18. Explain what count() does in terms of the dplyr verbs you just learned. What does the sort argument to count() do?

```
Answer
  flights |>
    count(origin, dest, sort = FALSE) # sort = FALSE by
                                                              default
# A tibble: 224 x 3
   origin dest
   <chr>
          <chr> <int>
 1 EWR
          ALB
                   439
 2 EWR
          ANC
                     8
 3 EWR
          ATL
                  5022
                   968
 4 EWR
          AUS
          AVL
 5 EWR
                   265
 6 EWR
          BDL
                  443
 7 EWR
          BNA
                  2336
  EWR
          BOS
                  5327
 9 EWR
          BQN
                   297
10 EWR
          BTV
                   931
# i 214 more rows
  flights |>
     count(origin, dest, sort = TRUE)
# A tibble: 224 x 3
   origin dest
                     n
   <chr>
          <chr> <int>
 1 JFK
          LAX
                 11262
 2 LGA
          ATL
                 10263
 3 LGA
          ORD
                  8857
 4 JFK
          SFO
                  8204
 5 LGA
          CLT
                  6168
 6 EWR
          ORD
                  6100
```

```
7 JFK BOS 5898
8 LGA MIA 5781
9 JFK MCO 5464
10 EWR BOS 5327
# i 214 more rows
Your text answer here.
```

19. Suppose we have the following tiny data frame:

```
df <- tibble(
    x = 1:5,
    y = c("a", "b", "a", "a", "b"),
    z = c("K", "K", "L", "L", "K")
)</pre>
```

a. Write down what you think the output will look like, then check if you were correct, and describe what group_by() does.

```
df |>
  group_by(y)
```

```
Answer
  df |>
     group_by(y)
# A tibble: 5 x 3
# Groups:
             y [2]
      х у
  <int> <chr> <chr>
1
      1 a
               K
2
      2 b
               K
3
      3 a
               L
4
      4 a
               L
5
      5 b
               K
Your text answer here.
```

b. Write down what you think the output will look like, then check if you were correct, and describe what arrange() does. Also comment on how it's different from the group_by() in part (a).

```
df |>
  arrange(y)
```

```
Answer
  df |>
     arrange(y)
# A tibble: 5 x 3
               Z
      х у
  <int> <chr> <chr>
               K
1
      1 a
2
      3 a
               L
3
               L
      2 b
               K
5
      5 b
Your text answer here.
```

c. Write down what you think the output will look like, then check if you were correct, and describe what the pipeline does.

```
df |>
  group_by(y) |>
  summarize(mean_x = mean(x))
```

d. Write down what you think the output will look like, then check if you were correct, and describe what the pipeline does. Then, comment on what the message says.

```
df |>
  group_by(y, z) |>
  summarize(mean_x = mean(x))
```

```
Answer
  df |>
    group_by(y, z) |>
    summarize(mean_x = mean(x))
`summarise()` has grouped output by 'y'. You can override using the `.groups`
argument.
# A tibble: 3 x 3
# Groups:
            y [2]
              mean_x
  <chr> <chr> <dbl>
        K
2 a
                 3.5
        L
3 b
        K
                 3.5
Your text answer here.
```

e. Write down what you think the output will look like, then check if you were correct, and describe what the pipeline does. How is the output different from the one in part (d)?

```
df |>
  group_by(y, z) |>
  summarize(mean_x = mean(x), .groups = "drop")
```

```
Answer
  df |>
    group_by(y, z) |>
    summarize(mean_x = mean(x), .groups = "drop")
# A tibble: 3 x 3
  У
        Z
  <chr> <chr> <dbl>
1 a
       K
2 a
       L
                 3.5
3 b
       K
                 3.5
```

f. Write down what you think the outputs will look like, then check if you were correct, and describe what each pipeline does. How are the outputs of the two pipelines different?

```
df |>
  group_by(y, z) |>
  summarize(mean_x = mean(x))

df |>
  group_by(y, z) |>
  mutate(mean_x = mean(x))
```

```
Answer
  df |>
    group_by(y, z) |>
    summarize(mean_x = mean(x))
`summarise()` has grouped output by 'y'. You can override using the `.groups`
argument.
# A tibble: 3 x 3
# Groups:
            y [2]
        z
              mean_x
  <chr> <chr>
               <dbl>
1 a
        K
                 1
2 a
        L
                 3.5
3 b
        K
                 3.5
  df |>
    group_by(y, z) |>
    mutate(mean_x = mean(x))
# A tibble: 5 x 4
# Groups:
            y, z [3]
      х у
                    mean_x
              Z
  <int> <chr> <chr>
                     <dbl>
              K
                        1
1
      1 a
2
      2 b
              K
                        3.5
3
      3 a
              L
                       3.5
4
              L
                        3.5
      4 a
```

5 5 b K 3.5

Your text answer here.

4 Exercises (Chapter 4)

1. Restyle the following pipelines following the guidelines above.

```
flights|>filter(dest=="IAH")|>group_by(year,month,day)|>summarize(n=n(),
    delay=mean(arr_delay,na.rm=TRUE))|>filter(n>10)

flights|>filter(carrier=="UA",dest%in%c("IAH","HOU"),sched_dep_time>
    0900,sched_arr_time<2000)|>group_by(flight)|>summarize(delay=mean(
    arr_delay,na.rm=TRUE),cancelled=sum(is.na(arr_delay)),n=n())|>filter(n>10)
```

Answer

Your R code here

Provide an easy way to restyle the code.

5 Exercises (Chapter 5)

```
Tables
  table1
# A tibble: 6 x 4
 country
               year cases population
  <chr>
              <dbl>
                    <dbl>
                                <dbl>
1 Afghanistan 1999
                     745
                             19987071
2 Afghanistan 2000
                      2666
                             20595360
3 Brazil
               1999 37737 172006362
4 Brazil
               2000 80488 174504898
5 China
               1999 212258 1272915272
6 China
               2000 213766 1280428583
  table2
# A tibble: 12 \times 4
   country
                year type
                                     count
   <chr>
               <dbl> <chr>
                                     <dbl>
 1 Afghanistan 1999 cases
                                       745
 2 Afghanistan 1999 population
                                  19987071
 3 Afghanistan 2000 cases
                                      2666
 4 Afghanistan 2000 population
                                  20595360
 5 Brazil
                1999 cases
                                     37737
 6 Brazil
                1999 population 172006362
 7 Brazil
                2000 cases
                                     80488
 8 Brazil
                2000 population
                                174504898
 9 China
                1999 cases
                                    212258
10 China
                1999 population 1272915272
11 China
                2000 cases
                                    213766
12 China
                2000 population 1280428583
  table3
```

1. For each of the sample tables, describe what each observation and each column represents.

```
Answer

Your text answer here.
```

- 2. Sketch out the process you'd use to calculate the rate for table2 and table3. You will need to perform four operations:
 - a. Extract the number of TB cases per country per year.
 - b. Extract the matching population per country per year.
 - c. Divide cases by population, and multiply by 10000.
 - d. Store back in the appropriate place.

You haven't yet learned all the functions you'd need to actually perform these operations, but you should still be able to think through the transformations you'd need.

```
Answer
  table2 |>
    pivot_wider(names_from = type,
                values from = count) |>
    mutate(rate = cases / population * 10000)
# A tibble: 6 x 5
  country
               year cases population rate
                                <dbl> <dbl>
  <chr>
              <dbl>
                     <dbl>
1 Afghanistan 1999
                       745
                             19987071 0.373
2 Afghanistan 2000
                      2666
                             20595360 1.29
3 Brazil
               1999 37737 172006362 2.19
4 Brazil
               2000 80488
                           174504898 4.61
5 China
               1999 212258 1272915272 1.67
6 China
               2000 213766 1280428583 1.67
```

```
#
  table3 |>
    separate_wider_delim(
      cols = rate,
      delim = "/",
      names = c("cases", "population"),
    ) |>
    mutate(
      cases = as.numeric(cases),
      population = as.numeric(population),
      rate = cases / population * 10000
# A tibble: 6 x 5
  country
               year cases population rate
  <chr>
              <dbl>
                     <dbl>
                                <dbl> <dbl>
1 Afghanistan 1999
                       745
                             19987071 0.373
2 Afghanistan
              2000
                      2666
                             20595360 1.29
3 Brazil
               1999 37737 172006362 2.19
4 Brazil
               2000 80488 174504898 4.61
5 China
               1999 212258 1272915272 1.67
               2000 213766 1280428583 1.67
6 China
```

For table2, we need to reshape the data to have a column for cases and a column for population and then divide the two to calculate the rate. A possible approach is shown above.

For table3, we need to separate cases and population into their own columns and then divide them. A possible approach is shown above.

6 Exercises (Chapter 6)

1. Go to the RStudio Tips Twitter account, https://twitter.com/rstudiotips and find one tip that looks interesting. Practice using it!

Answer

Your text answer here.

2. What other common mistakes will RStudio diagnostics report? Read https://support.posit.co/hc/en-us/articles/205753617-Code-Diagnostics to find out.

Answer

Your text answer here.

7 Exercises (Chapter 7)

1. What function would you use to read a file where fields were separated with "|"?

Answer

Your text answer here.

2. Apart from file, skip, and comment, what other arguments do read_csv() and read_tsv() have in common?

Answer

Your text answer here.

3. What are the most important arguments to read_fwf()?

Answer

Your text answer here.

4. Sometimes strings in a CSV file contain commas. To prevent them from causing problems, they need to be surrounded by a quoting character, like " or '. By default, read_csv() assumes that the quoting character will be ". To read the following text into a data frame, what argument to read_csv() do you need to specify?

```
"x,y\n1,'a,b'"
```

Answer

We need to specify the quote argument.

```
read_csv("x,y\n1,'a,b'", quote = "\'")
```

Rows: 1 Columns: 2

-- Column specification -----

Delimiter: ","
chr (1): y

5. Identify what is wrong with each of the following inline CSV files. What happens when you run the code?

```
read_csv("a,b\n1,2,3\n4,5,6")
read_csv("a,b,c\n1,2\n1,2,3,4")
read_csv("a,b\n\"1")
read_csv("a,b\n1,2\na,b")
read_csv("a;b\n1;3")
```

```
Answer
  read_csv("a,b\n1,2,3\n4,5,6")
Warning: One or more parsing issues, call `problems()` on your data frame for details,
e.g.:
  dat <- vroom(...)</pre>
  problems(dat)
Rows: 2 Columns: 2
-- Column specification -
Delimiter: ","
dbl (1): a
num (1): b
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.
# A tibble: 2 x 2
            b
  <dbl> <dbl>
      1
           23
      4
           56
There are only two column headers but three values in each row, so the last two
```

get merged.

```
Answer
  read csv("a,b,c\n1,2\n1,2,3,4")
Warning: One or more parsing issues, call `problems()` on your data frame for details,
 dat <- vroom(...)</pre>
 problems(dat)
Rows: 2 Columns: 3
-- Column specification ------
Delimiter: ","
dbl (2): a, b
num (1): c
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.
# A tibble: 2 x 3
           b
  <dbl> <dbl> <dbl>
1
      1
           2
                 NA
here are only three column headers, first row is missing a value in the last column
so gets an NA there, the second row has four values so the last two get merge
```

Answer

```
read_csv("a,b\n\"1")

Rows: 0 Columns: 2
-- Column specification -----
Delimiter: ","
chr (2): a, b

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.
# A tibble: 0 x 2
# i 2 variables: a <chr>, b <chr>
```

No rows are read in.

```
Answer
  read_csv("a,b\n1,2\na,b")
Rows: 2 Columns: 2
-- Column specification -----
Delimiter: ","
chr (2): a, b
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.
# A tibble: 2 x 2
  a
       b
  <chr> <chr>
       2
1 1
2 a
       b
Each column has a numerical and a character value, so the column type is coerced
to character.
```

```
Rows: 1 Columns: 1
-- Column specification -----
Delimiter: ","
chr (1): a;b

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.
# A tibble: 1 x 1
   `a;b`
   <chr>
1 1;3
The delimiter is; but it's not specified, therefore this is read in as a single-column data frame with a single observation.
```

6. Practice referring to non-syntactic names in the following data frame by:

```
set.seed(321)
annoying <- tibble(
   `1` = 1:10,
   `2` = `1` * 2 + rnorm(length(`1`))
)</pre>
```

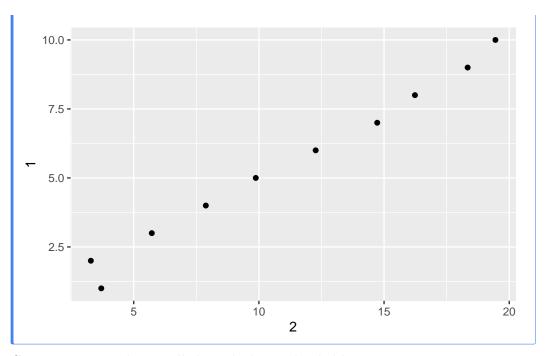
a. Extracting the variable called 1.

```
Answer
  annoying |>
    select(`1`)
# A tibble: 10 x 1
    `1`
  <int>
 1
      1
 2
      2
 3
      3
 4
      4
 5
     5
      6
 6
 7
     7
8
      8
      9
9
10
     10
  # or
  annoying$`1`
 [1] 1 2 3 4 5 6 7 8 9 10
```

b. Plotting a scatterplot of 1 vs. 2.

```
Answer

annoying |>
    ggplot(aes(x = `2`, y = `1`)) +
    geom_point()
```



c. Creating a new column called 3, which is 2 divided by 1.

```
Answer
  annoying |>
    mutate(^3 = ^2'/^1)
# A tibble: 10 \times 3
     `1`
           `2`
                  `3`
   <int> <dbl> <dbl>
          3.70
 1
                3.70
       2
         3.29
 2
                1.64
 3
       3 5.72 1.91
       4 7.88
                1.97
 4
 5
       5 9.88
                1.98
 6
       6 12.3
                 2.04
 7
       7 14.7
                2.10
 8
       8 16.2
                2.03
                 2.04
       9 18.3
 9
                1.94
      10 19.4
10
```

d. Renaming the columns to one, two, and three.

```
Answer
 annoying |>
   mutate(`3` = `2`/`1`) |>
     rename(
     "one" = `1`,
     "two" = ^2,
     "three" = 3
     )
# A tibble: 10 x 3
    one two three
  <int> <dbl> <dbl>
1 1 3.70 3.70
    2 3.29 1.64
3 3 5.72 1.91
  4 7.88 1.97
5 5 9.88 1.98
6 6 12.3 2.04
7 7 14.7 2.10
8 8 16.2 2.03
9
    9 18.3 2.04
10 10 19.4 1.94
```

8 Exercises (Chapter 9)

1. Create a scatterplot of hwy vs. displ where the points are pink filled in triangles.

```
Answer

# Your R code here
```

2. Why did the following code not result in a plot with blue points?

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

```
Answer

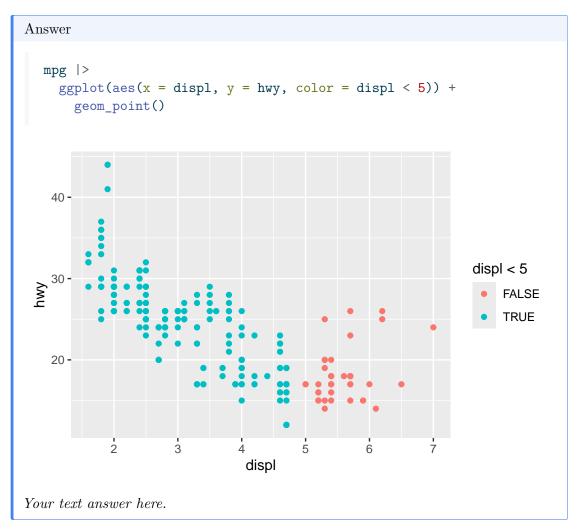
# Proper R code here

Your text answer here.
```

3. What does the stroke aesthetic do? What shapes does it work with? (Hint: use ?geom_point)

```
Answer
  mpg |>
    ggplot(aes(x = displ, y = hwy)) +
      geom_point(shape = 21, stroke = 0.5) -> p1
  mpg |>
    ggplot(aes(x = displ, y = hwy)) +
      geom_point(shape = 21, stroke = 1) -> p2
  mpg |>
    ggplot(aes(x = displ, y = hwy)) +
      geom_point(shape = 21, stroke = 2) -> p3
  library(patchwork)
  p1 / p2 / p3
                                    displ
                                    displ
                                    displ
Your text answer here.
```

4. What happens if you map an aesthetic to something other than a variable name, like aes(color = displ < 5)? Note, you'll also need to specify x and y.



5. What geom would you use to draw a line chart? A boxplot? A histogram? An area chart?

Answer Your text answer here. # R Code here

Answer

 $Your\ text\ answer\ here.$

R code here

Answer

Your text answer here.

R code here

Answer

Your text answer here.

Youe R code here

6. Earlier in this chapter we used show.legend without explaining it:

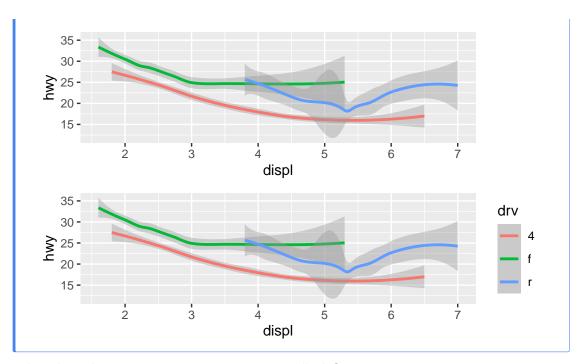
```
ggplot(mpg, aes(x = displ, y = hwy)) +
geom_smooth(aes(color = drv), show.legend = FALSE)
```

What does show.legend = FALSE do here? What happens if you remove it? Why do you think we used it earlier?

Answer

Your text answer here.

```
ggplot(mpg, aes(x = displ, y = hwy)) +
  geom_smooth(aes(color = drv), show.legend = FALSE) -> p1
ggplot(mpg, aes(x = displ, y = hwy)) +
  geom_smooth(aes(color = drv), show.legend = TRUE) -> p2
p1 / p2
```

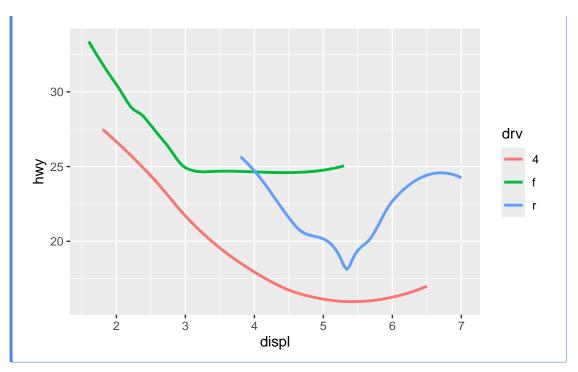


7. What does the se argument to geom_smooth() do?

```
Answer

Your text answer here.

ggplot(mpg, aes(x = displ, y = hwy, color = drv)) +
geom_smooth(se = FALSE)
```



8. Recreate the R code necessary to generate the following graphs. Note that wherever a categorical variable is used in the plot, it's drv.

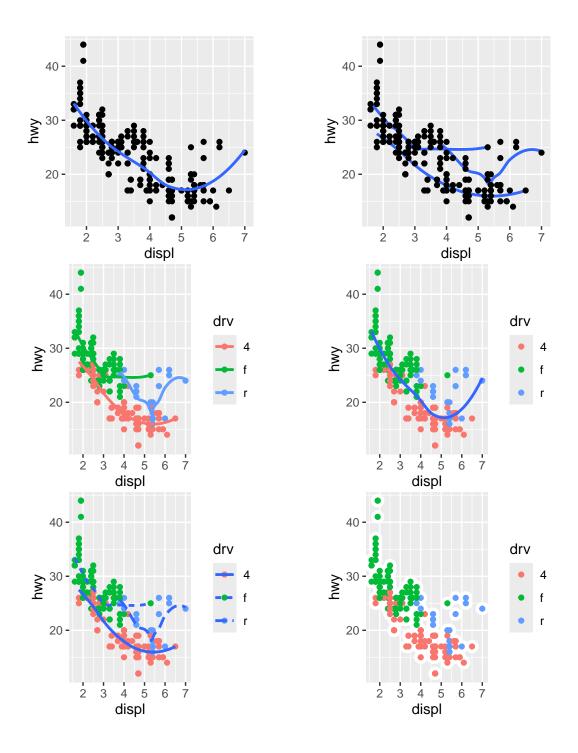
Answer The code for each of the plots is given below. # Your R code here

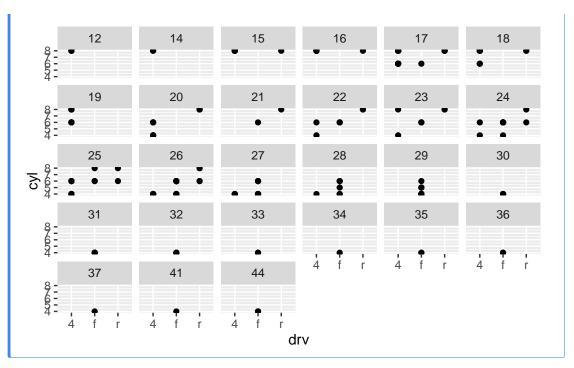
9. What happens if you facet on a continuous variable?

```
Answer

Your text answer here.

mpg |>
    ggplot(aes(x = drv, y = cyl)) +
    geom_point() +
    facet_wrap(~hwy)
```



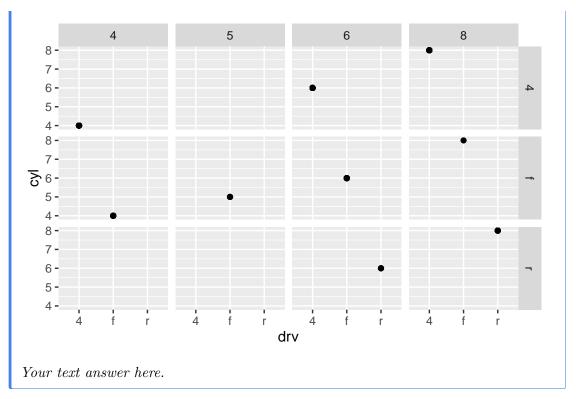


10. What do the empty cells in the plot above with facet_grid(drv ~ cyl) mean? Run the following code. How do they relate to the resulting plot?

```
ggplot(mpg) +
  geom_point(aes(x = drv, y = cyl))
```

```
Answer

ggplot(mpg) +
  geom_point(aes(x = drv, y = cyl)) +
  facet_grid(drv ~ cyl)
```



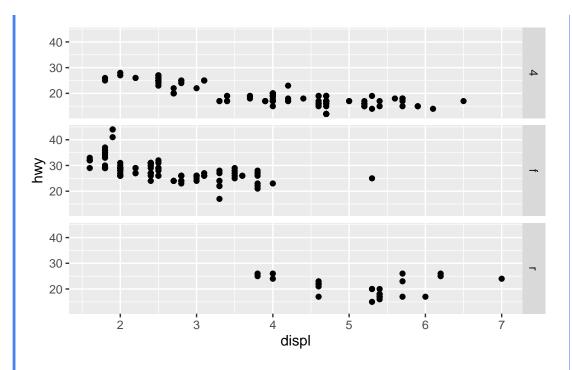
11. What plots does the following code make? What does . do?

```
ggplot(mpg) +
  geom_point(aes(x = displ, y = hwy)) +
  facet_grid(drv ~ .)

ggplot(mpg) +
  geom_point(aes(x = displ, y = hwy)) +
  facet_grid(. ~ cyl)
```

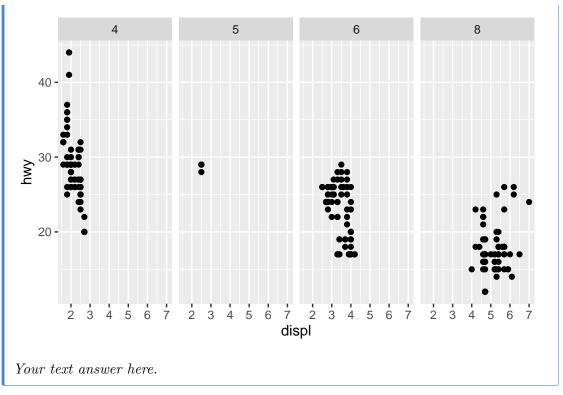
```
Answer

ggplot(mpg) +
  geom_point(aes(x = displ, y = hwy)) +
  facet_grid(drv ~ .)
```



Your text answer here.

```
ggplot(mpg) +
  geom_point(aes(x = displ, y = hwy)) +
  facet_grid(. ~ cyl)
```



12. Take the first faceted plot in this section:

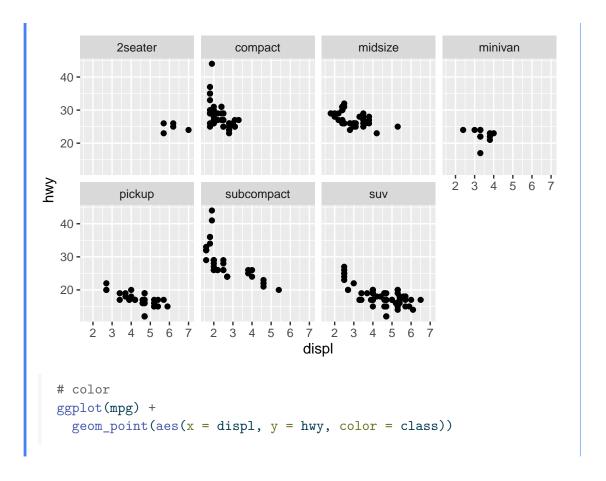
```
ggplot(mpg) +
  geom_point(aes(x = displ, y = hwy)) +
  facet_wrap(~ cyl, nrow = 2)
```

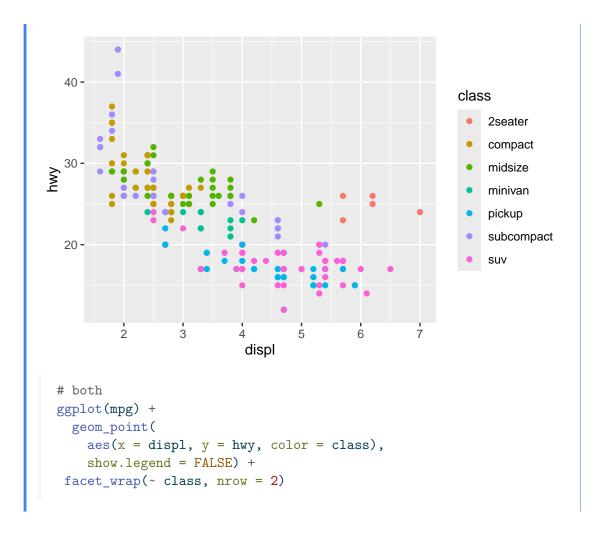
What are the advantages to using faceting instead of the color aesthetic? What are the disadvantages? How might the balance change if you had a larger dataset?

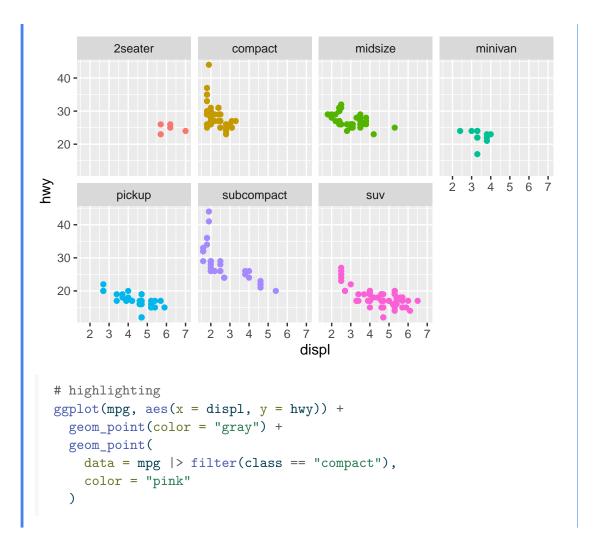
```
Answer

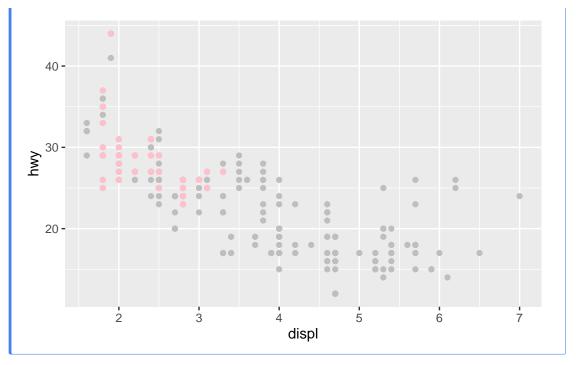
Your text answer here.

# facet
ggplot(mpg) +
geom_point(aes(x = displ, y = hwy)) +
facet_wrap(~ class, nrow = 2)
```









13. Read ?facet_wrap. What does nrow do? What does ncol do? What other options control the layout of the individual panels? Why doesn't facet_grid() have nrow and ncol arguments?

Answer

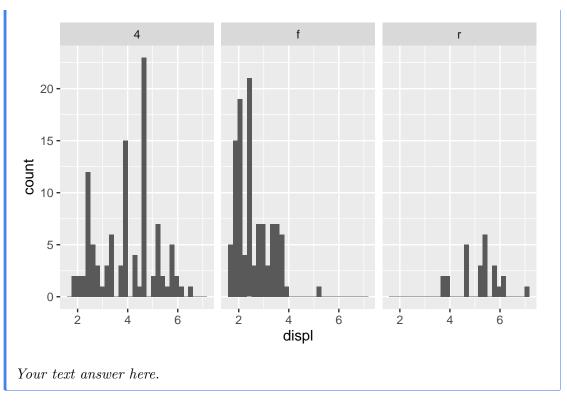
Your text answer here.

14. Which of the following plots makes it easier to compare engine size (displ) across cars with different drive trains? What does this say about when to place a faceting variable across rows or columns?

```
ggplot(mpg, aes(x = displ)) +
  geom_histogram() +
  facet_grid(drv ~ .)

ggplot(mpg, aes(x = displ)) +
  geom_histogram() +
  facet_grid(. ~ drv)
```

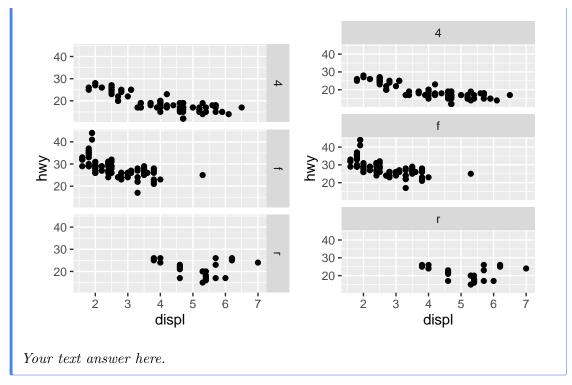
Answer ggplot(mpg, aes(x = displ)) +geom_histogram() + facet_grid(drv ~ .) `stat_bin()` using `bins = 30`. Pick better value with `binwidth`. 20 -15 -10 -5 -0 -20 oonut 15 -5 -0 -20 -15 **-**10-5 -0 -2 displ ggplot(mpg, aes(x = displ)) +geom_histogram() + facet_grid(. ~ drv) `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.



15. Recreate the following plot using facet_wrap() instead of facet_grid(). How do the positions of the facet labels change?

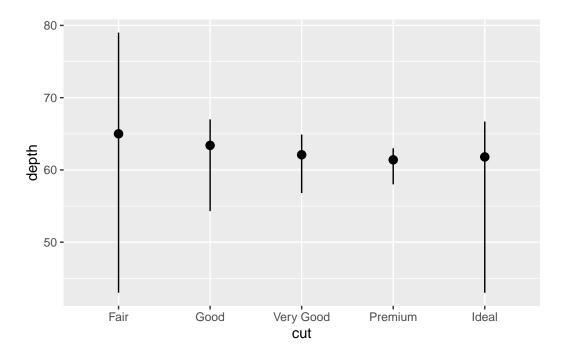
```
ggplot(mpg) +
  geom_point(aes(x = displ, y = hwy)) +
  facet_grid(drv ~ .)
```

```
ggplot(mpg) +
    geom_point(aes(x = displ, y = hwy)) +
    facet_grid(drv ~ .) -> p1
    ggplot(mpg) +
        geom_point(aes(x = displ, y = hwy)) +
        facet_wrap(~drv, nrow = 3) -> p2
    p1 + p2
```



16. What is the default geom associated with stat_summary()? How could you rewrite the previous plot to use that geom function instead of the stat function?

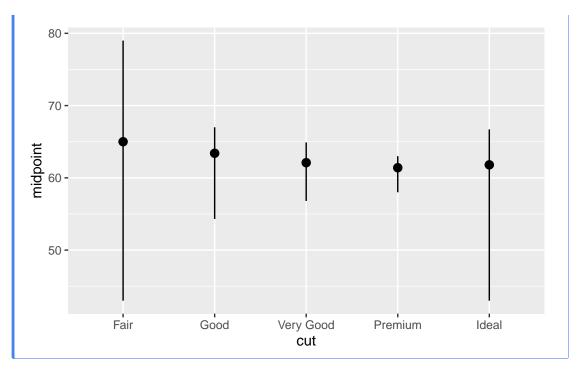
```
ggplot(diamonds) +
stat_summary(
  aes(x = cut, y = depth),
  fun.min = min,
  fun.max = max,
  fun = median
)
```



```
Answer

Your text answer here.

diamonds |>
    group_by(cut) |>
    summarize(
    lower = min(depth),
    upper = max(depth),
    midpoint = median(depth)
    ) |>
    ggplot(aes(x = cut, y = midpoint)) +
    geom_pointrange(aes(ymin = lower, ymax = upper))
```



17. What does geom_col() do? How is it different from geom_bar()?

Answer

Your text answer here.

18. Most geoms and stats come in pairs that are almost always used in concert. Make a list of all the pairs. What do they have in common? (Hint: Read through the documentation.)

Answer

Geoms and stats that are almost always used in concert are listed below:

stat
stat_count()
stat_bin_2d()
<pre>stat_boxplot()</pre>
stat_contour_filled()
stat_contour()
stat_sum()
stat_density_2d()
<pre>stat_density()</pre>
<pre>stat_bindot()</pre>

geom	stat
<pre>geom_function()</pre>	stat_function()
<pre>geom_sf()</pre>	stat_sf()
<pre>geom_sf()</pre>	stat_sf()
<pre>geom_smooth()</pre>	stat_smooth()
<pre>geom_violin()</pre>	<pre>stat_ydensity()</pre>
<pre>geom_hex()</pre>	<pre>stat_bin_hex()</pre>
<pre>geom_qq_line()</pre>	stat_qq_line()
geom_qq()	stat_qq()
<pre>geom_quantile()</pre>	stat_quantile()

19. What variables does stat_smooth() compute? What arguments control its behavior?

Answer

stat_smooth() computes the following variables:

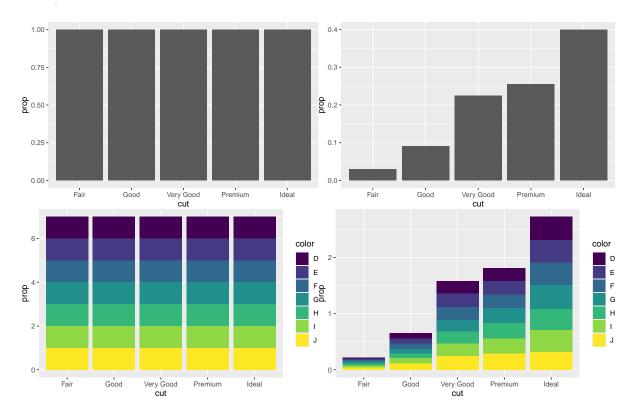
- y or x: Predicted value
- ymin or xmin: Lower pointwise confidence interval around the mean
- ymax or xmax: Upper pointwise confidence interval around the mean
- se: Standard error
- 20. In our proportion bar chart, we needed to set group = 1. Why? In other words, what is the problem with these two graphs?

```
ggplot(diamonds, aes(x = cut, y = after_stat(prop))) +
   geom_bar()

ggplot(diamonds, aes(x = cut, fill = color, y = after_stat(prop))) +
   geom_bar()
```

Answer

Your text answer here.



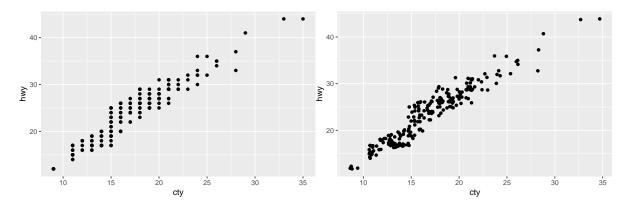
21. What is the problem with the following plot? How could you improve it?

```
ggplot(mpg, aes(x = cty, y = hwy)) +
  geom_point()
```

Answer

Your text answer here.

```
ggplot(mpg, aes(x = cty, y = hwy)) +
  geom_point()
ggplot(mpg, aes(x = cty, y = hwy)) +
  geom_jitter()
```



22. What, if anything, is the difference between the two plots? Why?

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

ggplot(mpg, aes(x = displ, y = hwy)) +
  geom_point(position = "identity")
```

Answer

Your text answer here.

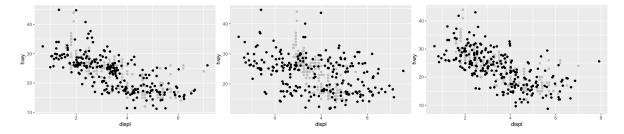
Your R code here

23. What parameters to geom_jitter() control the amount of jittering?

Answer

Your text answer here.

```
set.seed(321)
ggplot(mpg, aes(x = displ, y = hwy)) +
   geom_point(color = "gray") +
   geom_jitter(height = 1, width = 1)
ggplot(mpg, aes(x = displ, y = hwy)) +
   geom_point(color = "gray") +
   geom_jitter(height = 1, width = 5)
ggplot(mpg, aes(x = displ, y = hwy)) +
   geom_point(color = "gray") +
   geom_point(color = "gray") +
   geom_jitter(height = 5, width = 1)
```

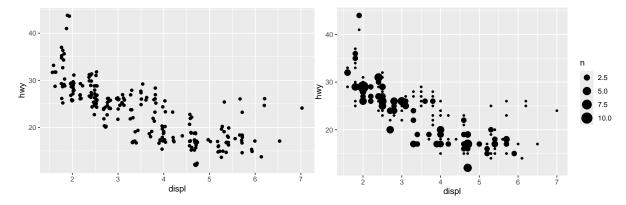


24. Compare and contrast geom_jitter() with geom_count().

Answer

Your text answer here.

```
ggplot(mpg, aes(x = displ, y = hwy)) +
  geom_jitter()
ggplot(mpg, aes(x = displ, y = hwy)) +
  geom_count()
```



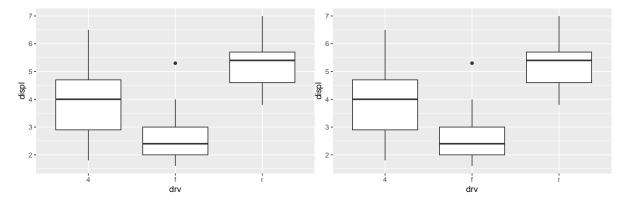
25. What's the default position adjustment for geom_boxplot()? Create a visualization of

the mpg dataset that demonstrates it.

Answer

Your text answer here.

```
ggplot(mpg, aes(x = drv, y = displ)) +
  geom_boxplot()
ggplot(mpg, aes(x = drv, y = displ)) +
  geom_boxplot(position = "dodge2")
```



26. Turn a stacked bar chart into a pie chart using coord_polar().

Answer

Your text answer here.

Your R code here

27. What's the difference between coord_quickmap() and coord_map()?

Answer

Your text answer here.

28. What does the following plot tell you about the relationship between city and highway mpg? Why is coord_fixed() important? What does geom_abline() do?

```
ggplot(data = mpg, mapping = aes(x = cty, y = hwy)) +
  geom_point() +
  geom_abline() +
```

coord_fixed()