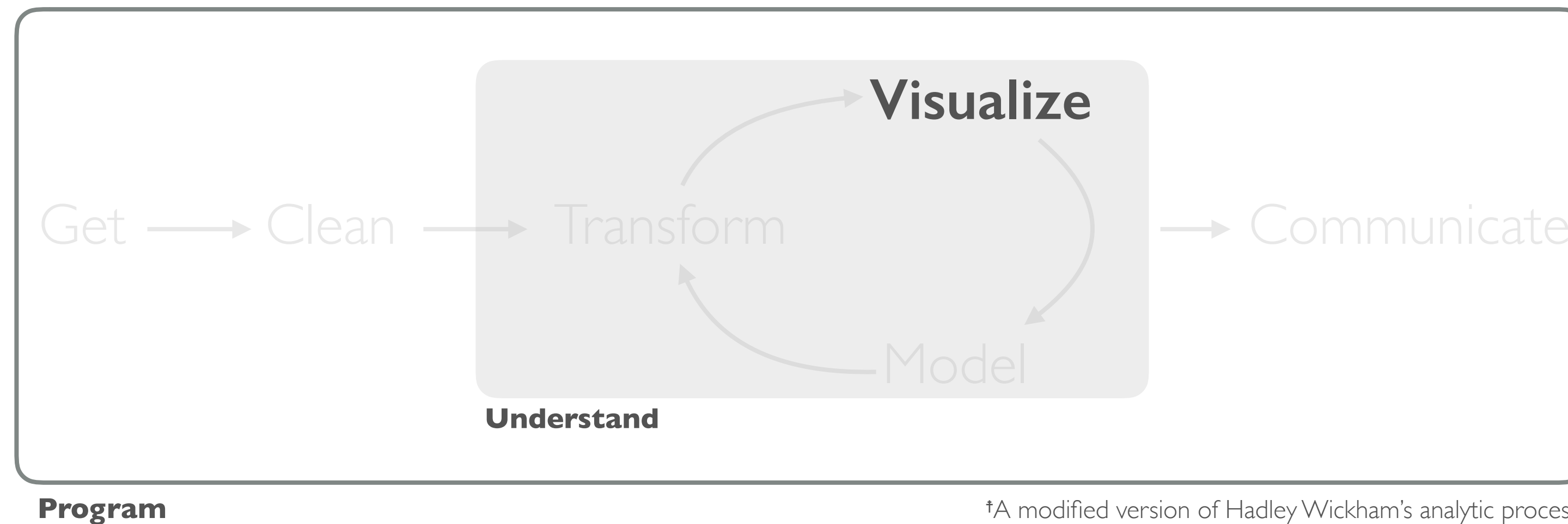


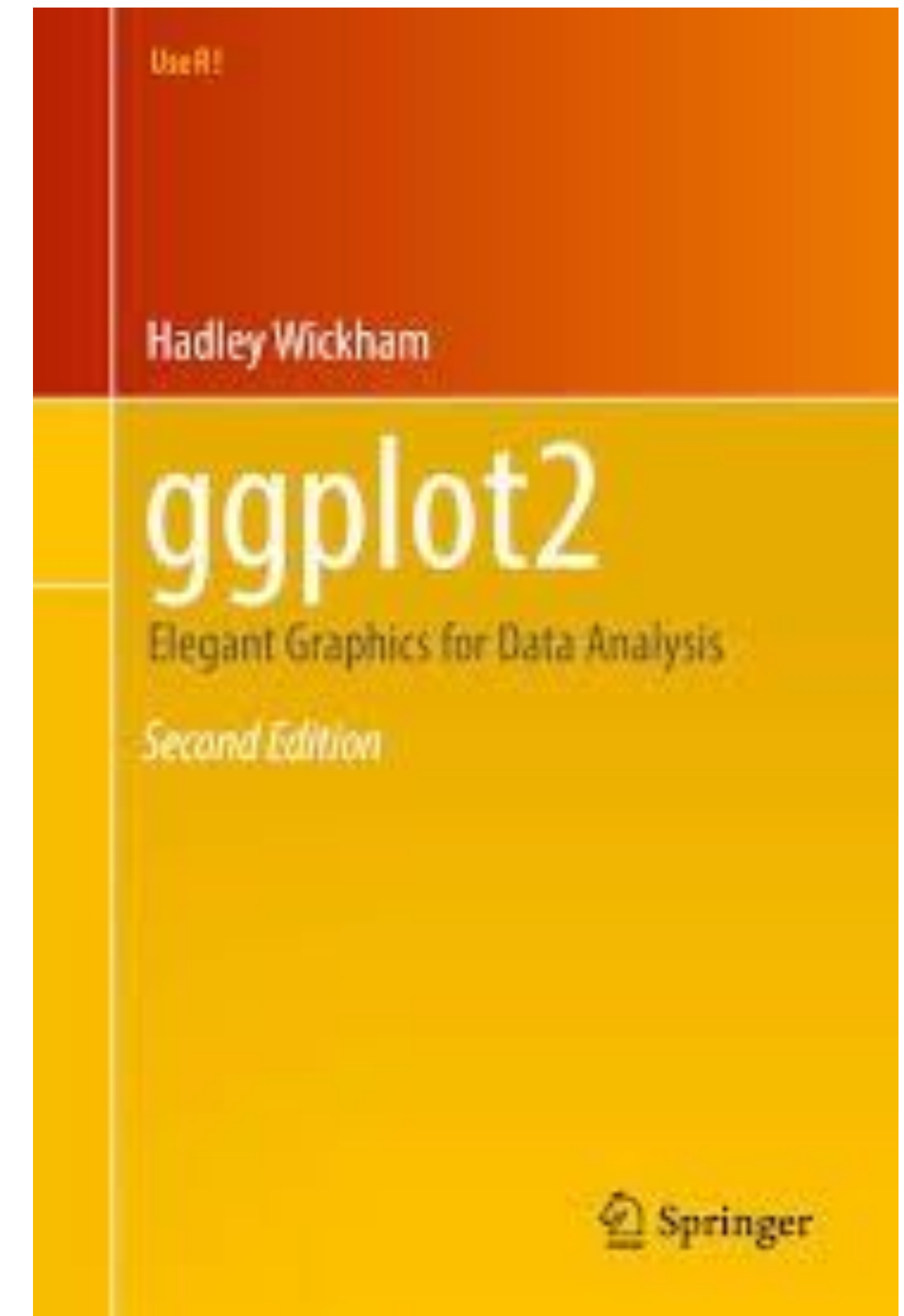
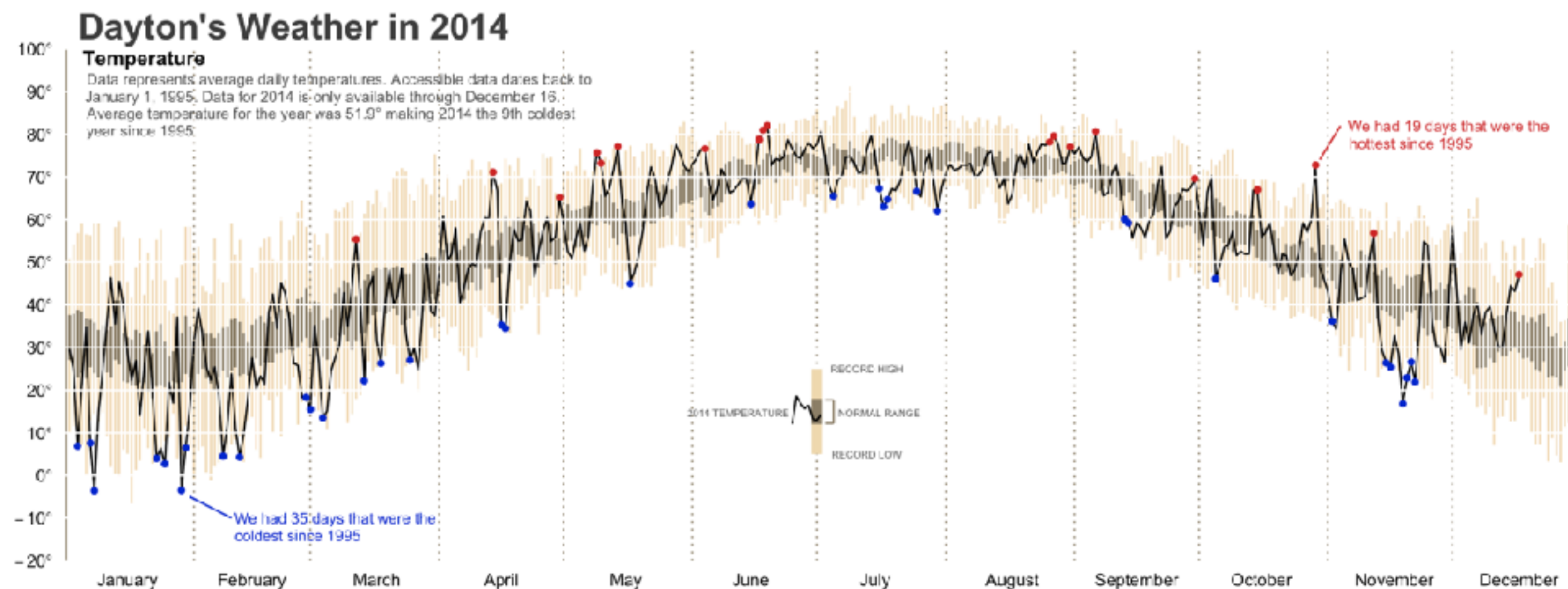
VISUALIZATION



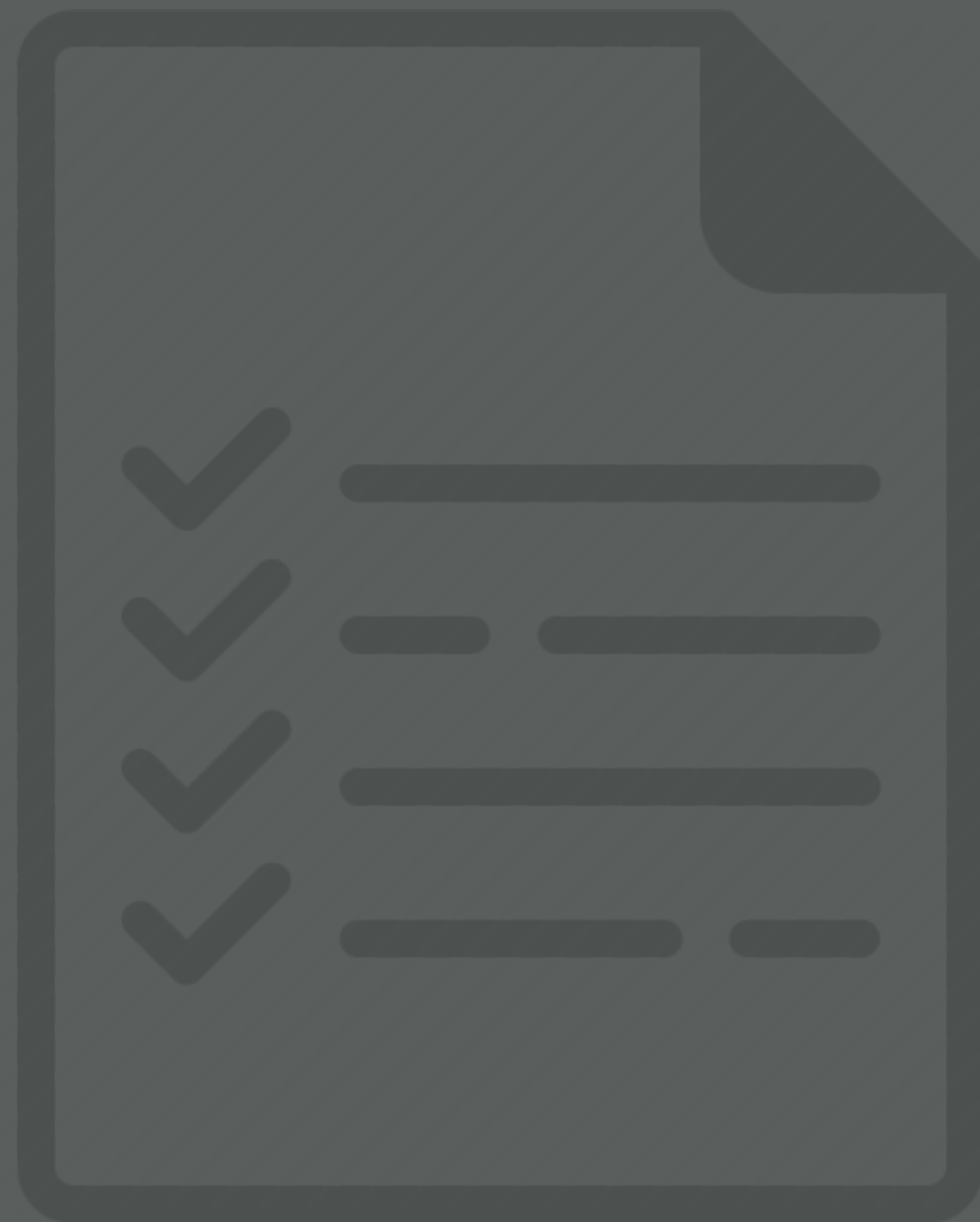
†A modified version of Hadley Wickham's analytic process

ggplot2

- R has several systems for making graphs
- `ggplot2` is the most elegant and versatile
- Implements the grammar of graphics theory behind data visualization



PREREQUISITES



PACKAGE PREREQUISITE

```
library(tidyverse)
#> Loading tidyverse: ggplot2
#> Loading tidyverse: tibble
#> Loading tidyverse: tidyr
#> Loading tidyverse: readr
#> Loading tidyverse: purrr
#> Loading tidyverse: dplyr
#> Conflicts with tidy packages -----
#> filter(): dplyr, stats
#> lag():    dplyr, stats
```

DATA PREREQUISITE

```
mpg
#> # A tibble: 234 × 11
#>   manufacturer model displ  year  cyl    trans  drv   cty   hwy fl
#>   <chr>    <chr> <dbl> <int> <int>    <chr> <chr> <int> <int> <chr>
#> 1      audi    a4   1.8  1999     4 auto(l5)  f    18    29   p
#> 2      audi    a4   1.8  1999     4 manual(m5)  f    21    29   p
#> 3      audi    a4   2.0  2008     4 manual(m6)  f    20    31   p
#> 4      audi    a4   2.0  2008     4  auto(av)  f    21    30   p
#> 5      audi    a4   2.8  1999     6 auto(l5)  f    16    26   p
#> 6      audi    a4   2.8  1999     6 manual(m5)  f    18    26   p
#> # ... with 228 more rows, and 1 more variables: class <chr>
```

Type ***View(mpg)*** in the console for a spreadsheet view of the data

YOUR TURN!

*Can you find the help file explaining the **mpg** data set?*

SOLUTION

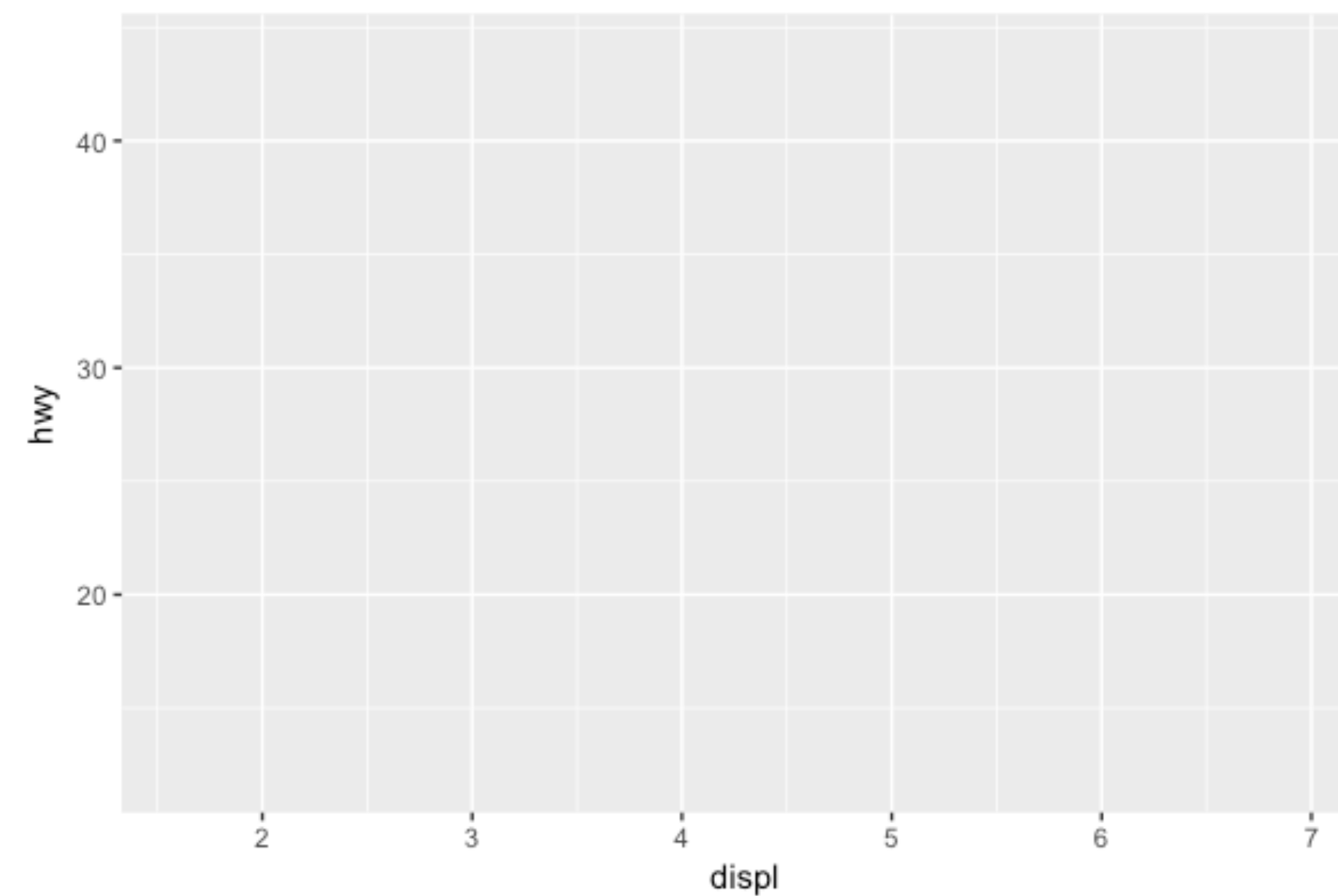
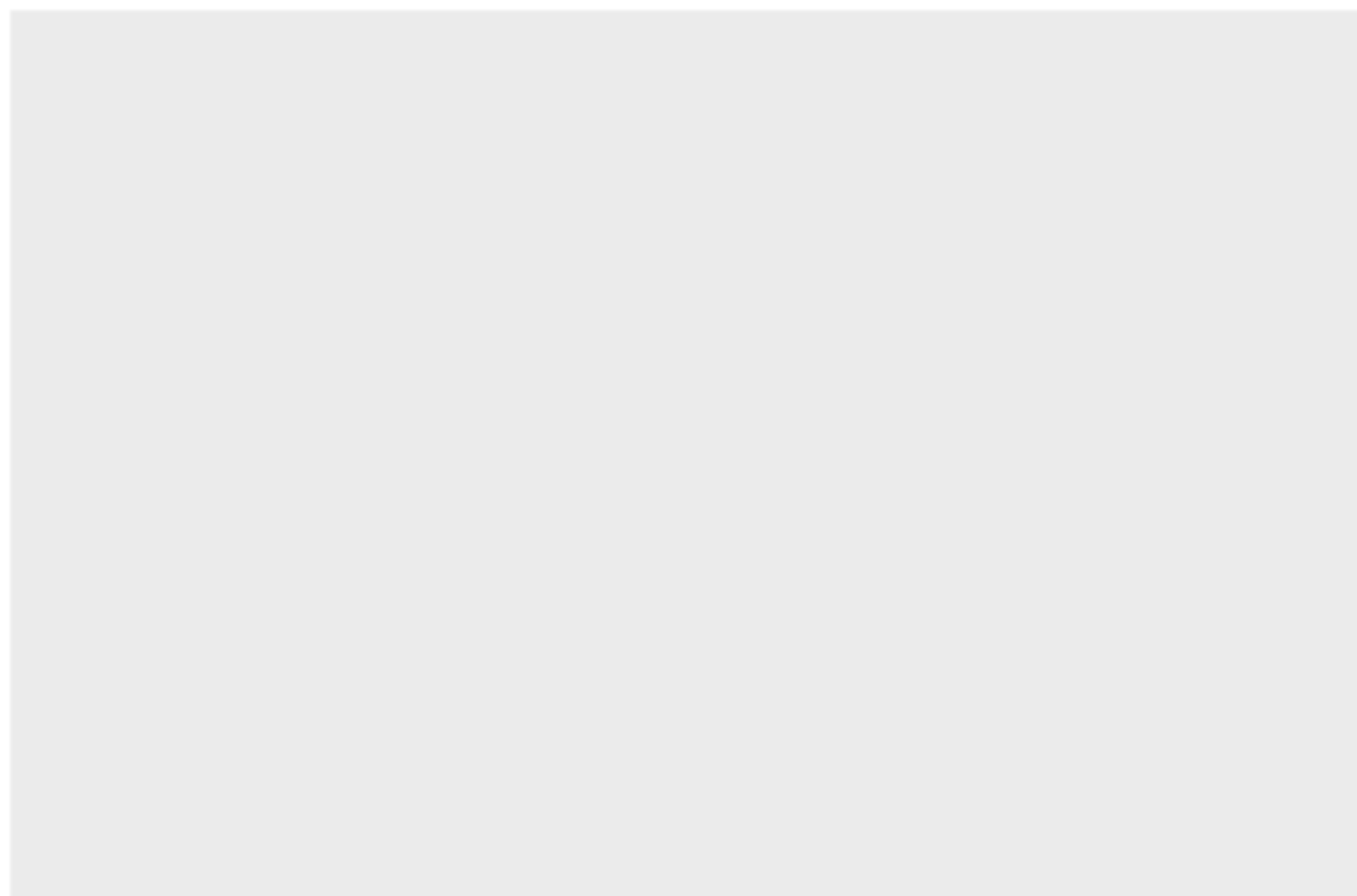
```
# additional documentation for the mpg data set  
?mpg
```


CANVAS

LET'S CREATE OUR "CANVAS"

```
# left  
ggplot(data = mpg)
```

```
# right  
ggplot(data = mpg, aes(x = displ, y = hwy))
```



GEOMS

geom_abline	geom_histogram
geom_bar	geom_jitter
geom_bin2d	geom_label
geom_blank	geom_map
geom_boxplot	geom_path
geom_contour	geom_point
geom_count	geom_polygon
geom_hex	geom_quantile
geom_crossbar	geom_raster
geom_density	geom_ribbon
geom_density_2d	geom_rug
geom_dotplot	geom_segment
geom_errorbarh	geom_smooth
geom_freqpoly	geom_violin

LETS ADD “GEOMS”

- We display data with geometric shapes
- ~ 30 built-in geoms (with many more offered by other pkgs)

Type **geom_** + *tab in the console*

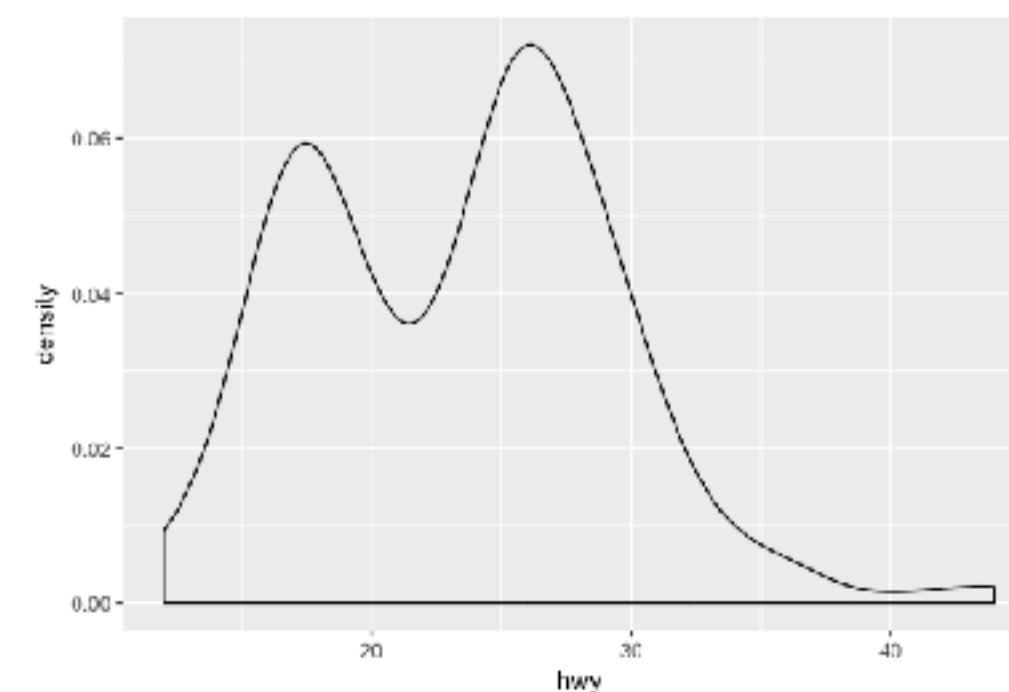
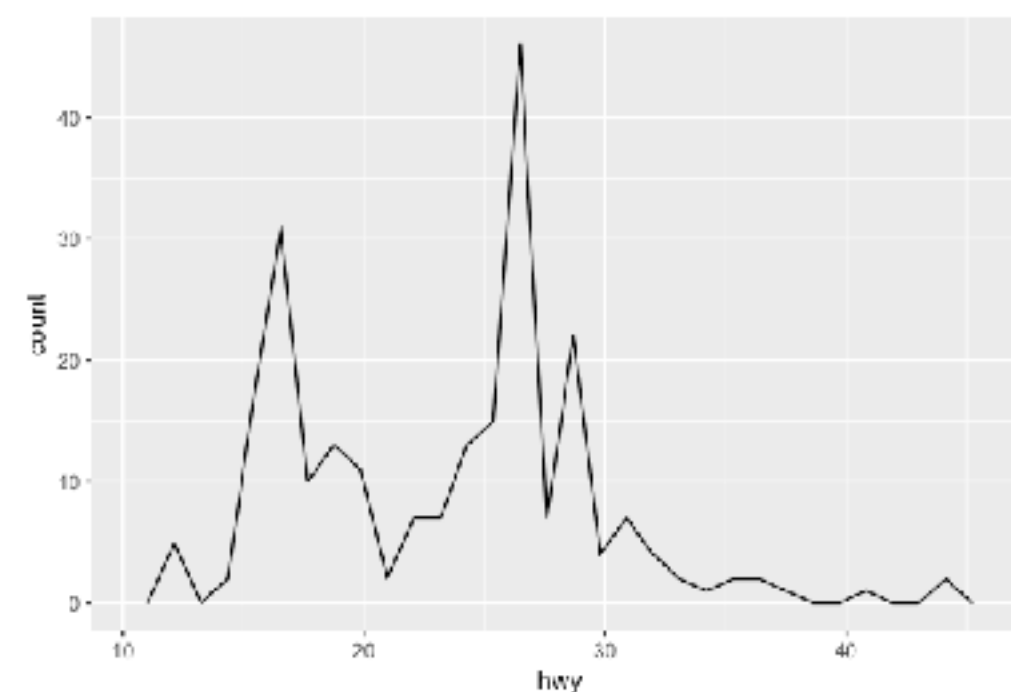
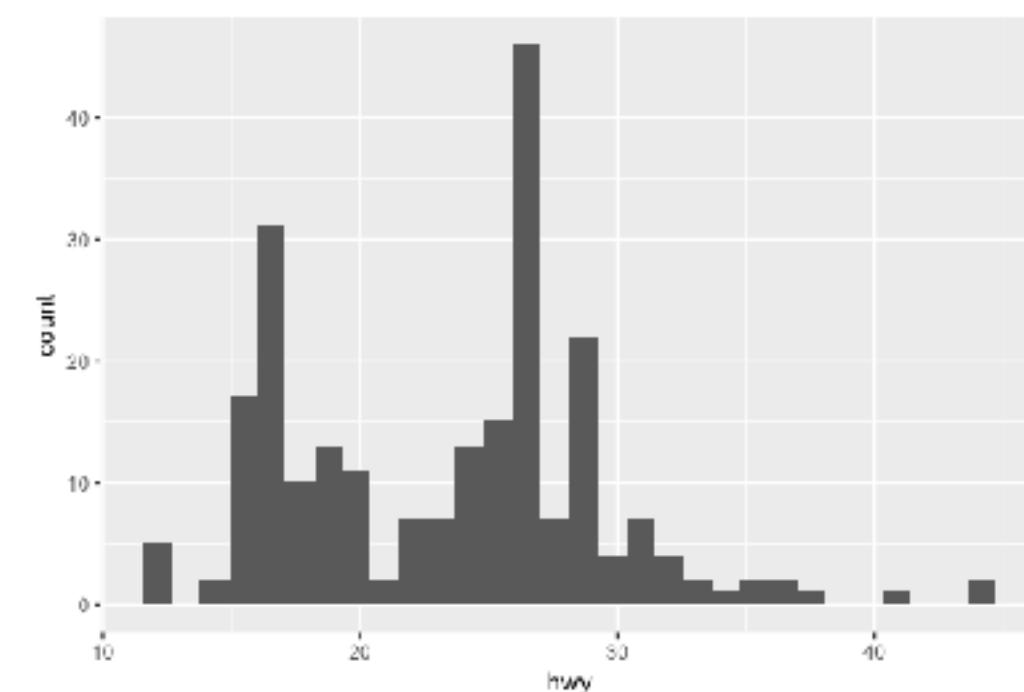
geom_abline	geom_histogram
geom_bar	geom_jitter
geom_bin2d	geom_label
geom_blank	geom_map
geom_boxplot	geom_path
geom_contour	geom_point
geom_count	geom_polygon
geom_hex	geom_quantile
geom_crossbar	geom_raster
geom_density	geom_ribbon
geom_density_2d	geom_rug
geom_dotplot	geom_segment
geom_errorbarh	geom_smooth
geom_freqpoly	geom_violin

UNIVARIATE GEOMS

```
ggplot(data = mpg, aes(x = hwy)) +  
  geom_histogram()
```

```
ggplot(data = mpg, aes(x = hwy)) +  
  geom_freqpoly()
```

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



geom_abline

geom_histogram

geom_bar

geom_jitter

geom_bin2d

geom_label

geom_blank

geom_map

geom_boxplot

geom_path

geom_contour

geom_point

geom_count

geom_polygon

geom_hex

geom_quantile

geom_crossbar

geom_raster

geom_density

geom_ribbon

geom_density_2d

geom_rug

geom_dotplot

geom_segment

geom_errorbarh

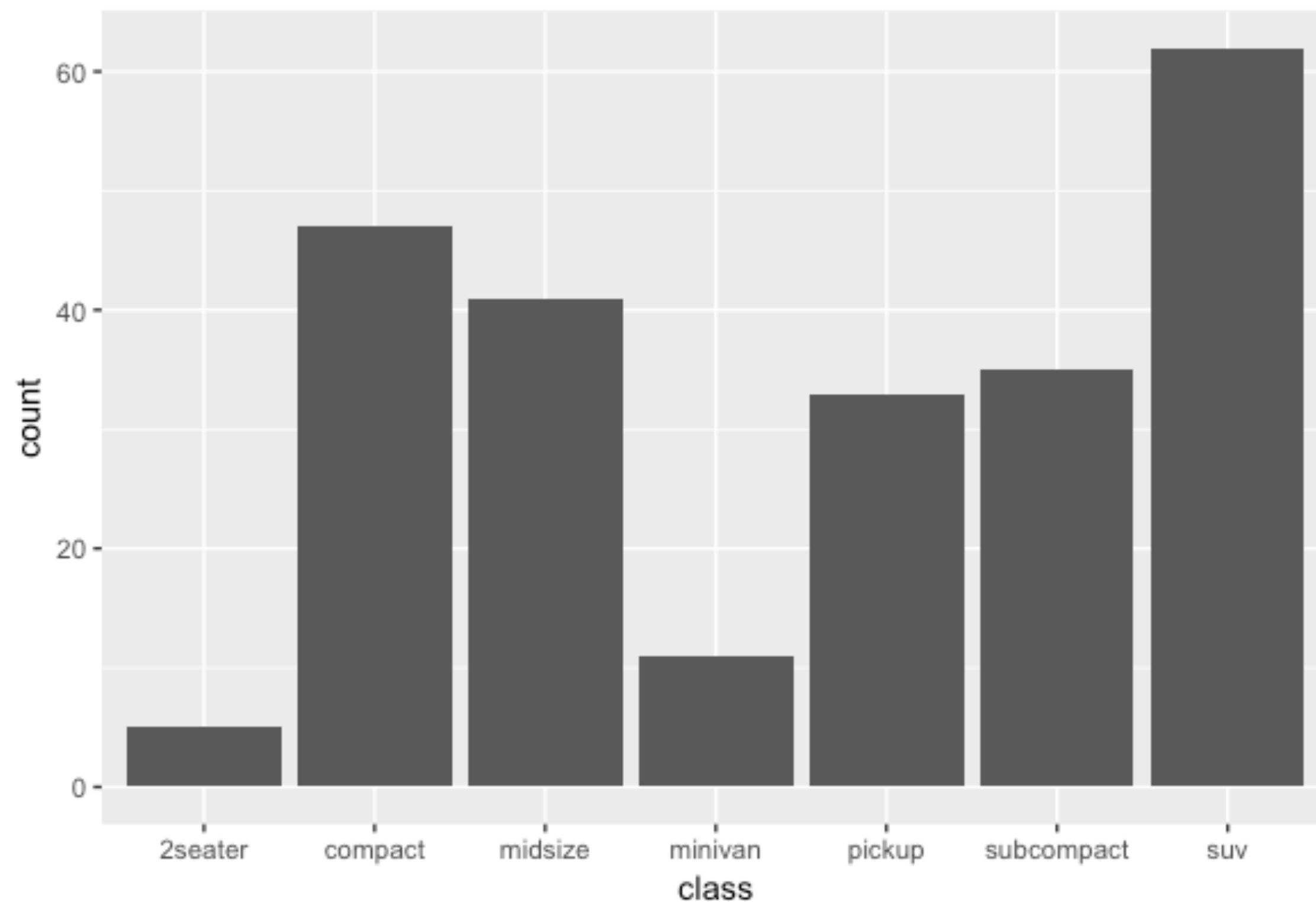
geom_smooth

geom_freqpoly

geom_violin

UNIVARIATE GEOMS

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



geom_abline

geom_histogram

geom_bar

geom_jitter

geom_bin2d

geom_label

geom_blank

geom_map

geom_boxplot

geom_path

geom_contour

geom_point

geom_count

geom_polygon

geom_hex

geom_quantile

geom_crossbar

geom_raster

geom_density

geom_ribbon

geom_density_2d

geom_rug

geom_dotplot

geom_segment

geom_errorbarh

geom_smooth

geom_freqpoly

geom_violin

UNIVARIATE GEOMS

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

- This is called an **aes**thetic mapping argument
- Every geom requires a mapping argument
 - Some geoms require just one (x variable)
 - While other geoms require two (x & y variable)

geom_abline

geom_histogram

geom_bar

geom_jitter

geom_bin2d

geom_label

geom_blank

geom_map

geom_boxplot

geom_path

geom_contour

geom_point

geom_count

geom_polygon

geom_hex

geom_quantile

geom_crossbar

geom_raster

geom_density

geom_ribbon

geom_density_2d

geom_rug

geom_dotplot

geom_segment

geom_errorbarh

geom_smooth

geom_freqpoly

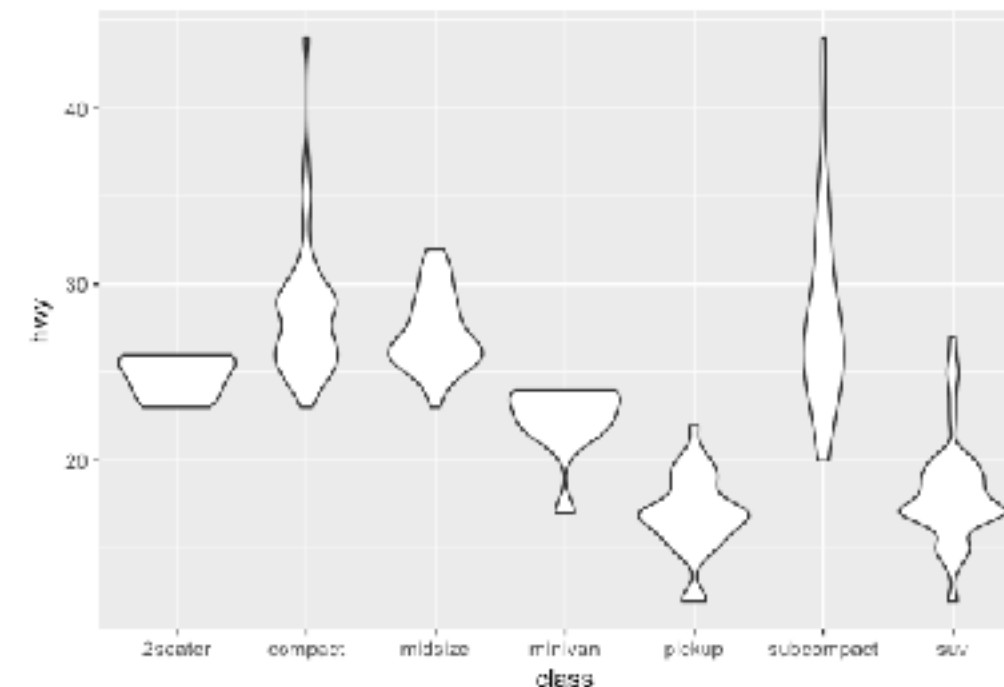
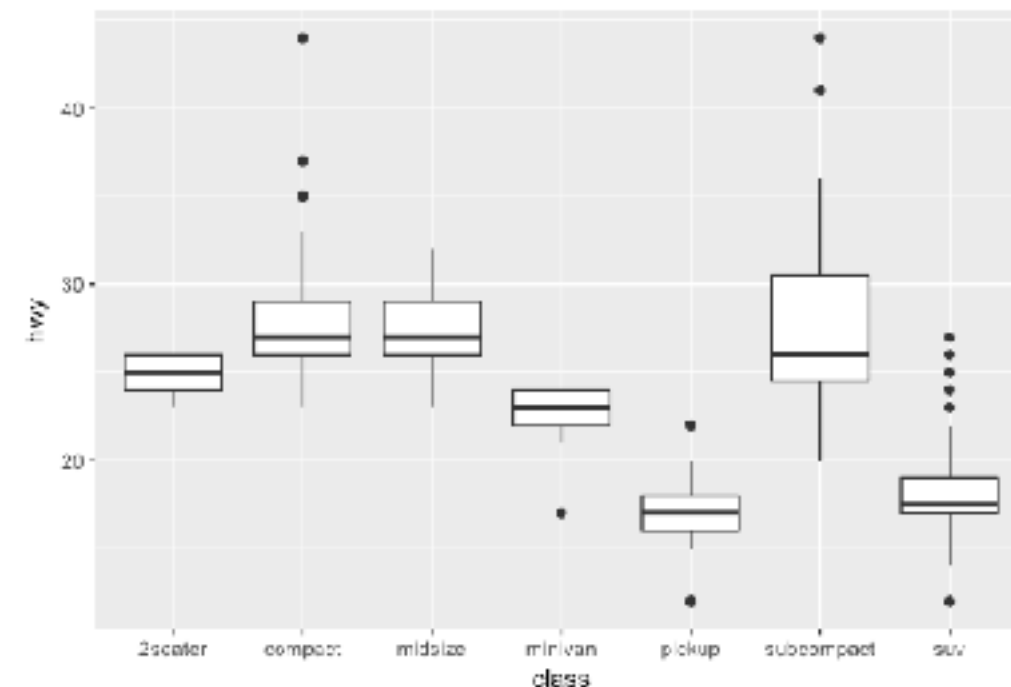
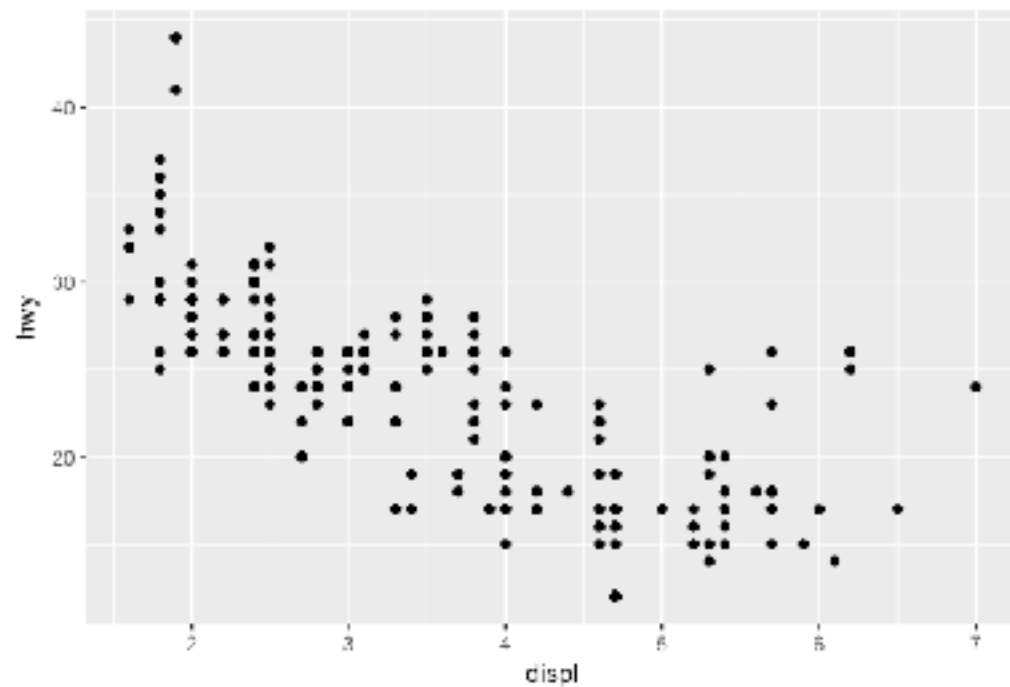
geom_violin

BIVARIATE GEOMS

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

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

```
ggplot(data = mpg, aes(x = class, y = hwy)) +  
  geom_violin()
```



geom_abline

geom_bar

geom_bin2d

geom_blank

geom_boxplot

geom_contour

geom_count

geom_hex

geom_crossbar

geom_density

geom_density_2d

geom_dotplot

geom_errorbarh

geom_freqpoly

geom_histogram

geom_jitter

geom_label

geom_map

geom_path

geom_point

geom_polygon

geom_quantile

geom_raster

geom_ribbon

geom_rug

geom_segment

geom_smooth

geom_violin

YOUR TURN!

1. Import the `CustomerData.csv` file
2. Create a chart that illustrates the distribution of the **DebtToIncomeRatio** variable.
3. Create a chart that shows the counts for each **JobCategory**
4. Create a scatter plot of **HHIncome** vs **CardSpendMonth**

SOLUTION

```
#1: import data
customer <- read_csv("data/CustomerData.csv")

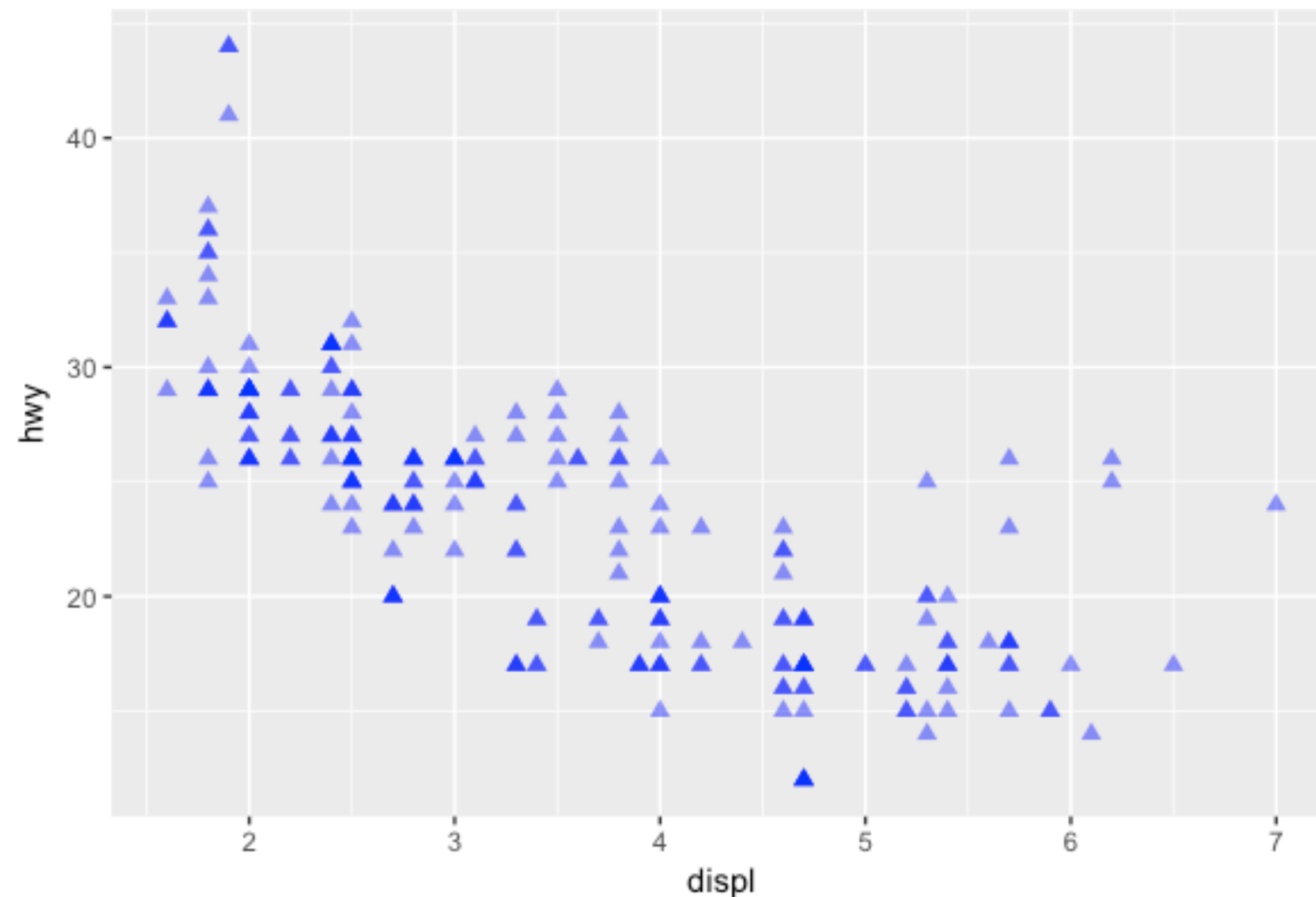
#2: distribution of DebtToIncomeRatio variable
ggplot(data = customer, aes(x = DebtToIncomeRatio)) +
  geom_histogram()

#3: distribution of JobCategory variable
ggplot(data = customer, aes(x = JobCategory)) +
  geom_bar()

#4: scatter plot for HHIncome vs CardSpendMonth
ggplot(data = customer, aes(x = HHIncome, y = CardSpendMonth)) +
  geom_point()
```

NON-MAPPING AESTHETICS

```
ggplot(data = mpg, aes(x = displ, y = hwy)) +  
  geom_point(color = "blue", size = 2, shape = 17, alpha = .5)
```



We can also change other visual aesthetics in our graphics

- color
- size
- shape (0-25 ?pch)
- opacity

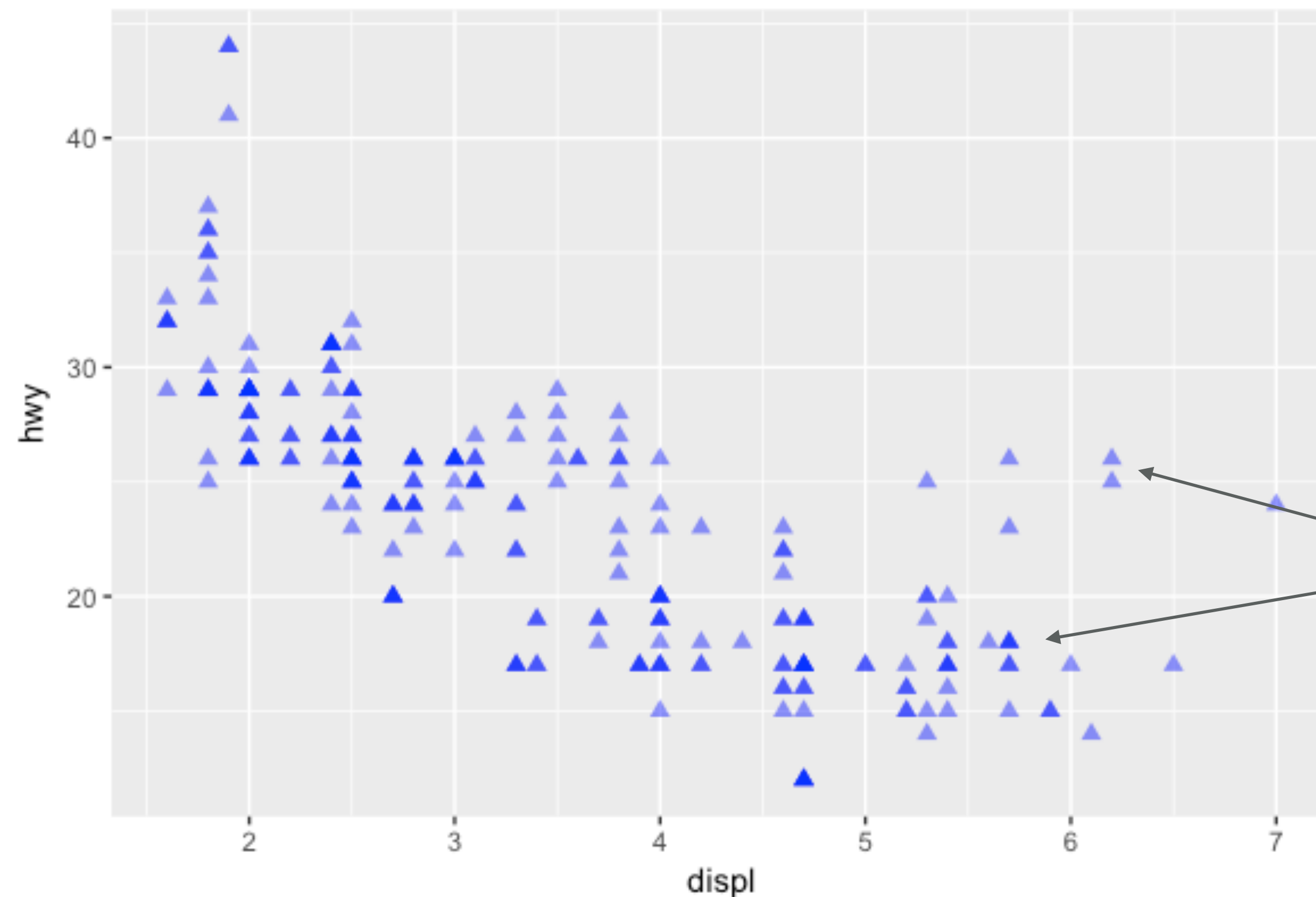
Lots of color and shape options; just google and you'll find plenty of references

NON-MAPPING AESTHETICS

```
ggplot(data = mpg, aes(x = displ, y = hwy)) +  
  geom_point(color = "blue", size = 2, shape = 17, alpha = .5)
```

We can also change other visual aesthetics in our graphics

- color
- size
- shape
- opacity



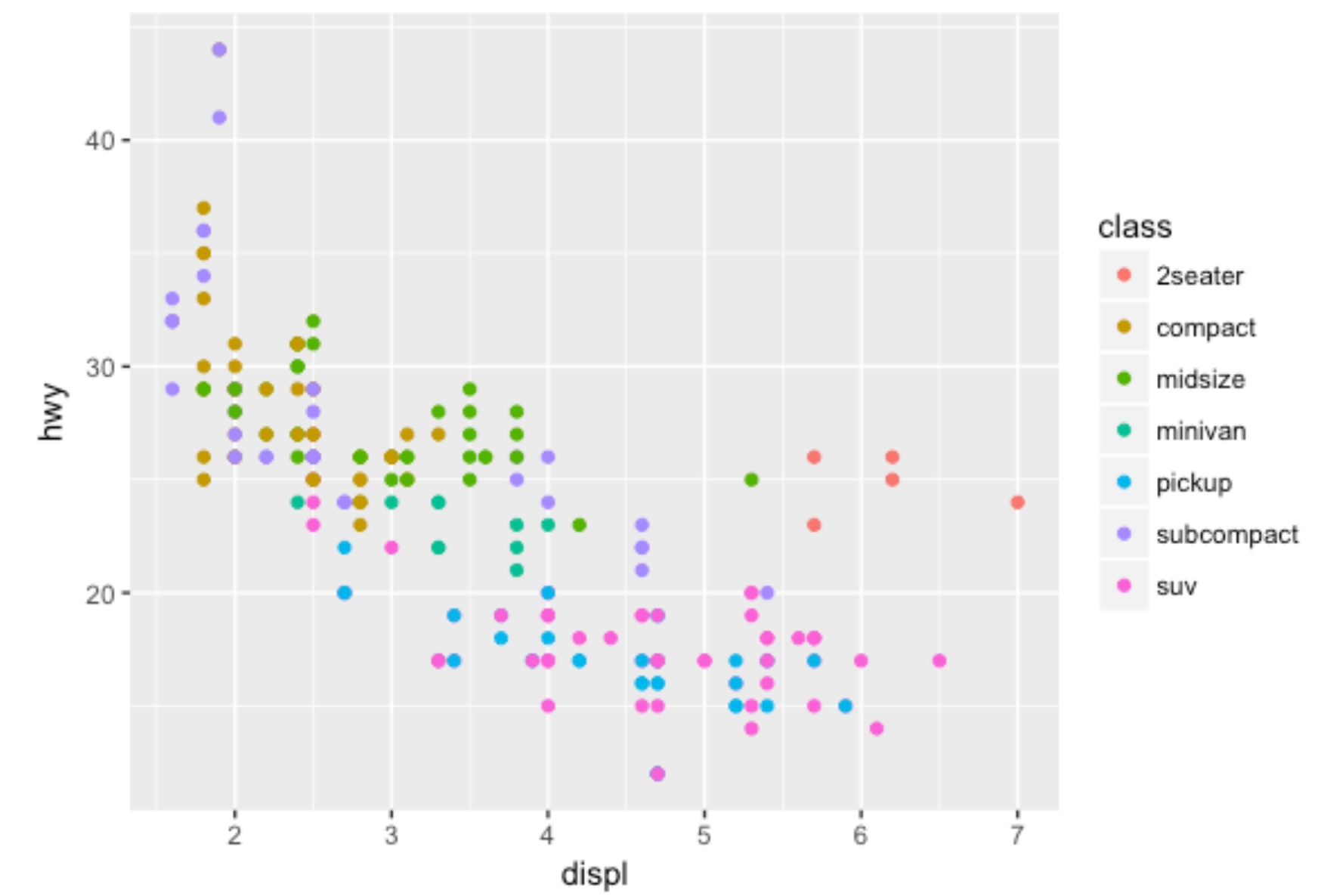
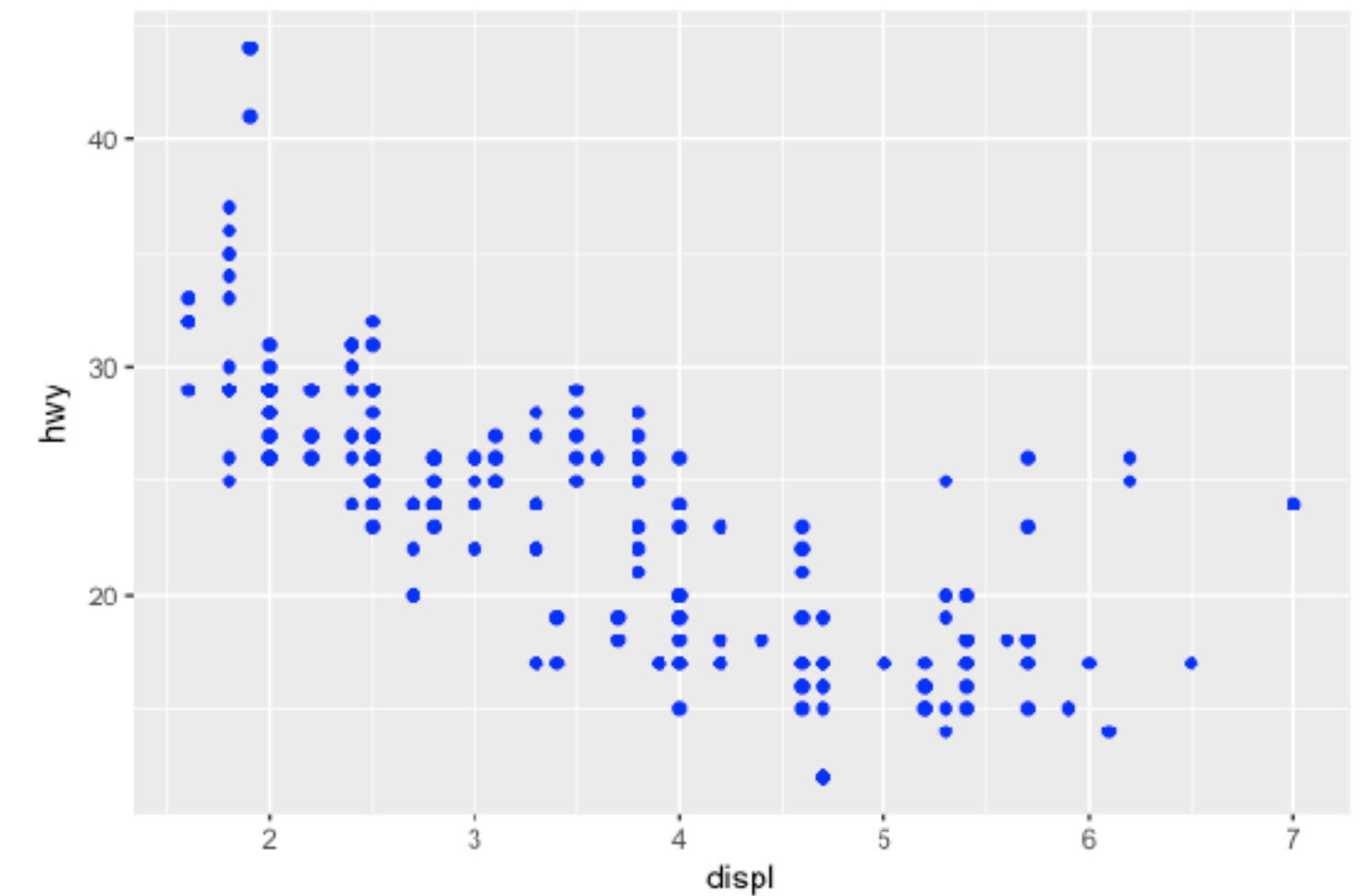
- Why are some points darker than others?
- Try **geom_jitter** in place of **geom_point**
- What do you think **geom_jitter** does?

ADDING A 3RD DIMENSION

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

By moving the `color` argument to within `aes()`, we can map a 3rd variable to our plot

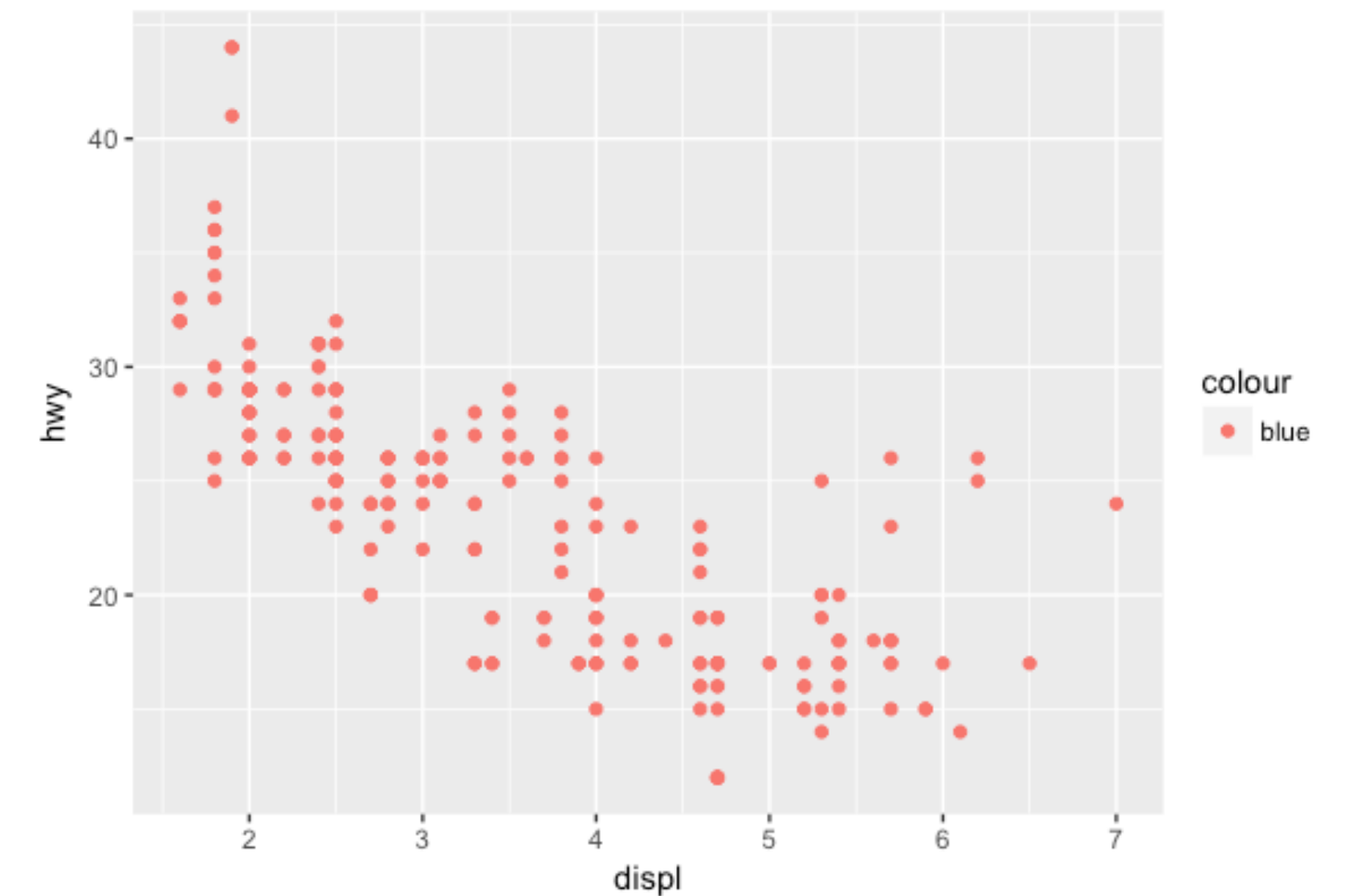
```
ggplot(data = mpg, aes(x = displ, y = hwy, color = class)) +  
  geom_point()
```



ADDING A 3RD DIMENSION

A common error...what happened???

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



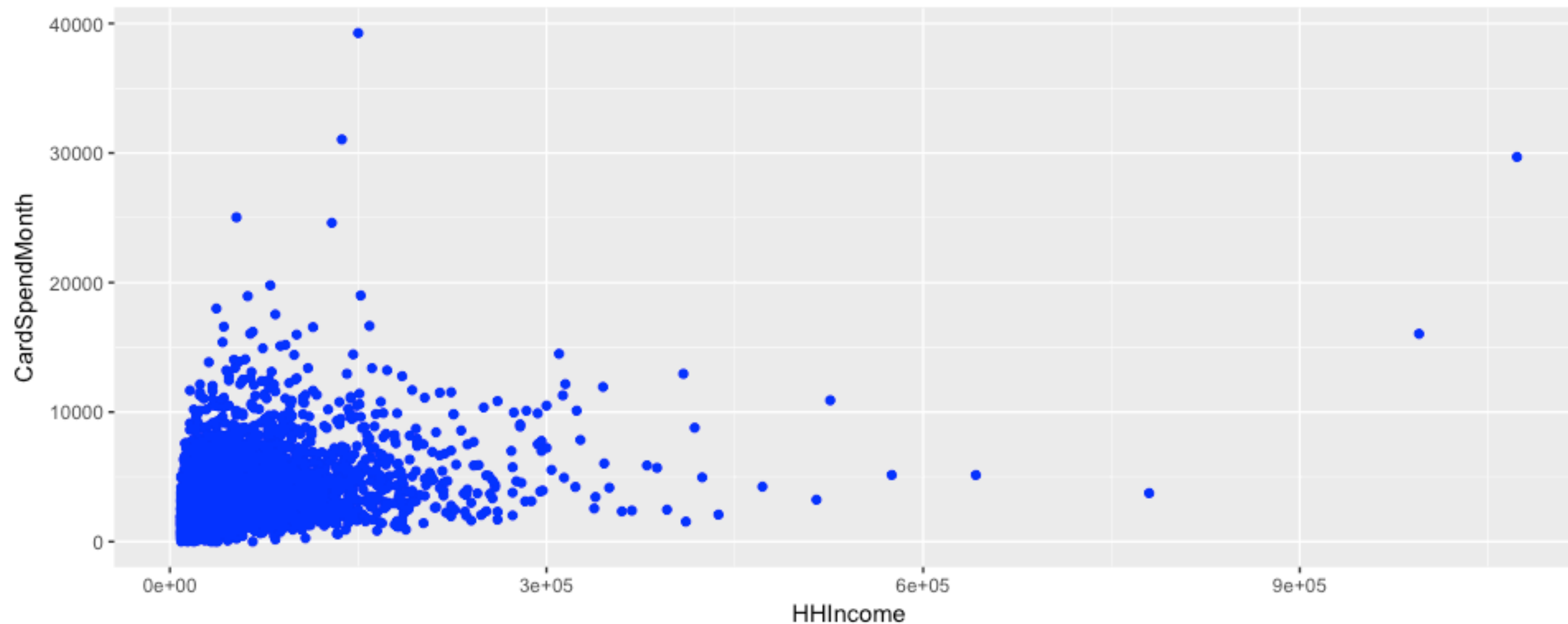
YOUR TURN!

1. Create a scatter plot of **HHIncome** vs **CardSpendMonth** and color all points blue.
2. Create a scatter plot of **HHIncome** vs **CardSpendMonth** and color all points based on whether or not the customer is retired.

SOLUTION

#1. Create a scatter plot of HHIncome vs CardSpendMonth and color all points blue.

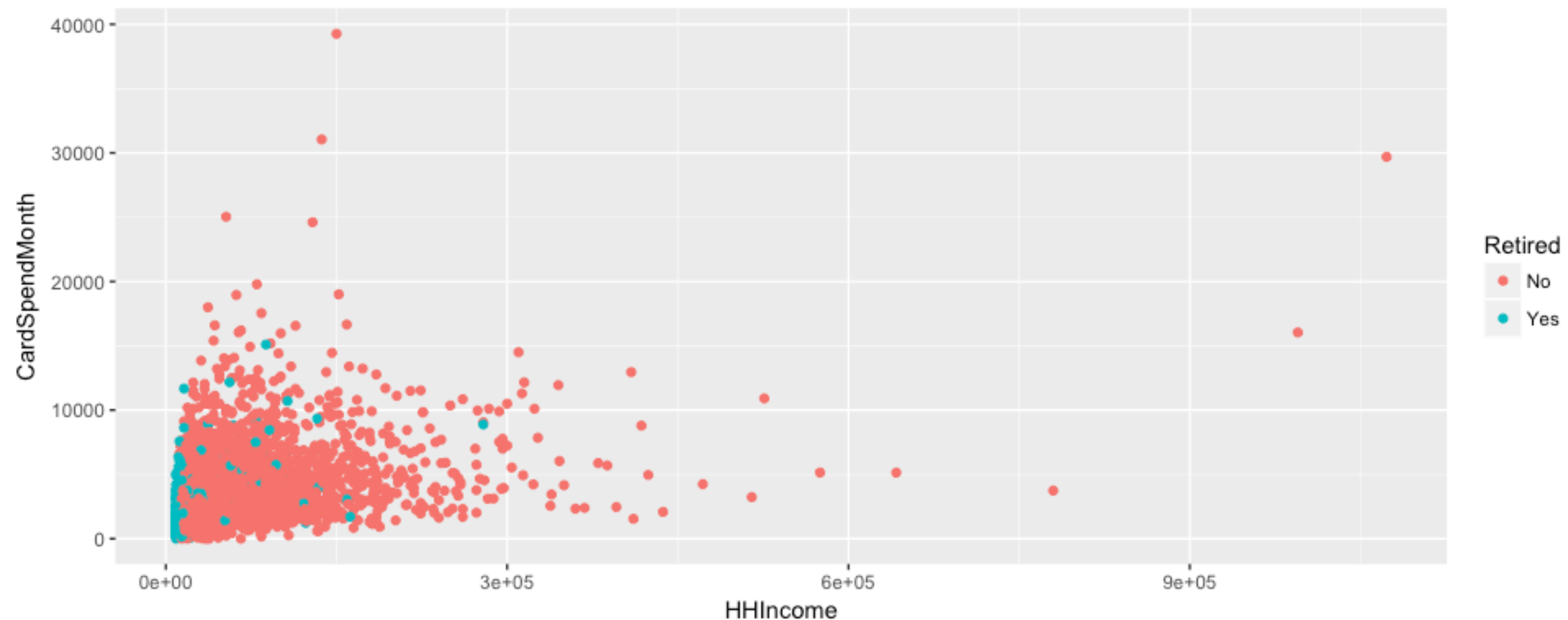
```
ggplot(customer, aes(x = HHIncome, y = CardSpendMonth)) +  
  geom_point(color = "blue")
```



SOLUTION

#2. Create a scatter plot of HHIncome vs CardSpendMonth and color all points based on whether or not the customer is retired..

```
ggplot(customer, aes(x = HHIncome, y = CardSpendMonth, color = Retired)) +  
  geom_point()
```





1964

1965

1966

1967

1968

1969

1970

1971

1972

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1974

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1978

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1980

1981

1982

1983

1984

1985

1986

1987

1988

1989

1990

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1998

1999

2000

2001

2002

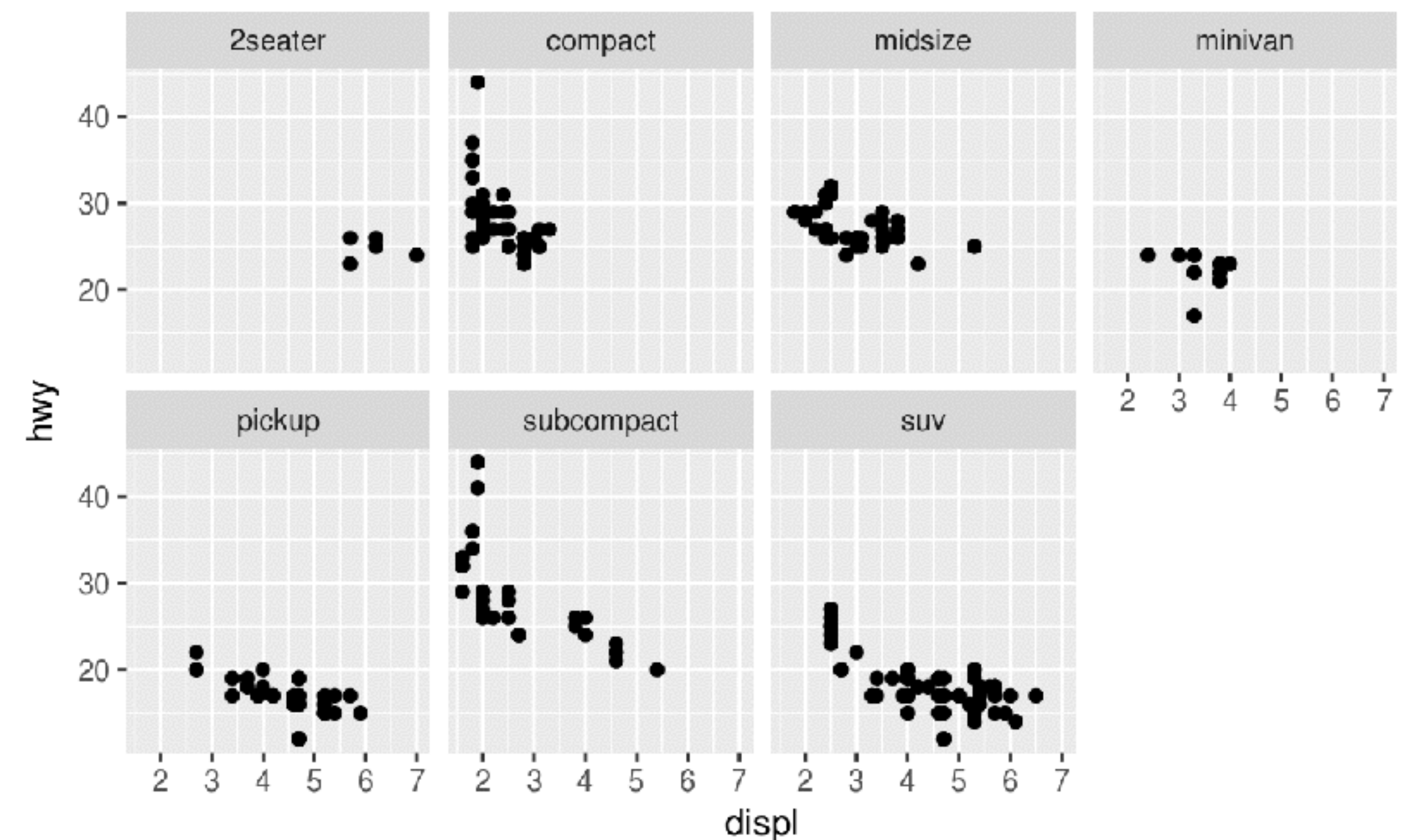
2003

FACETS

FACETS = SMALL MULTIPLES

- The **facet** functions provide a simple way to create small multiples
 - **facet_wrap**: primarily used to create small multiples based on a single variable
 - **facet_grid**: primarily used to create a small multiples grid based on two variables

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

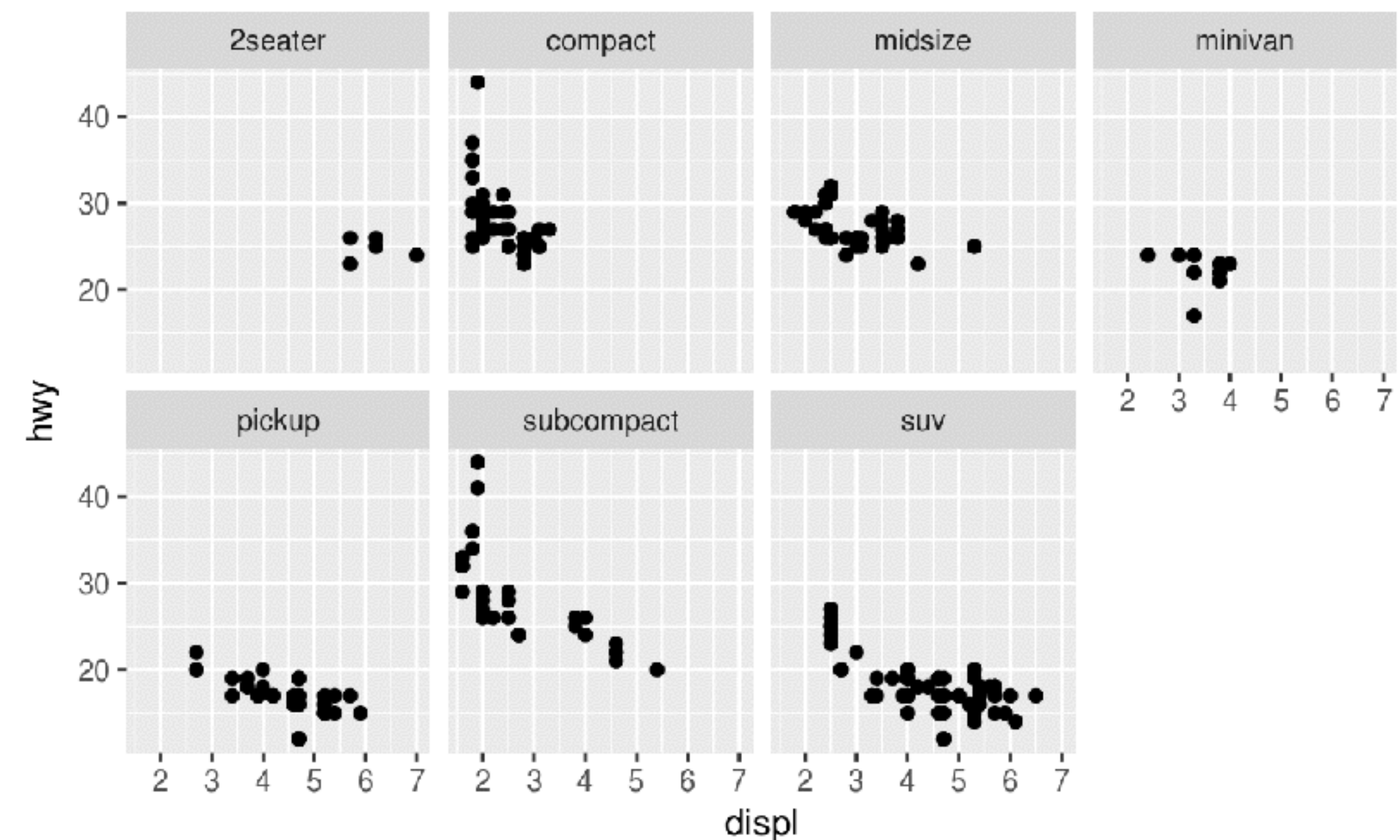


FACETS = SMALL MULTIPLES

- The **facet** functions provide a simple way to create small multiples
 - **facet_wrap**: primarily used to create small multiples based on a single variable
 - **facet_grid**: primarily used to create a small multiples grid based on two variables

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

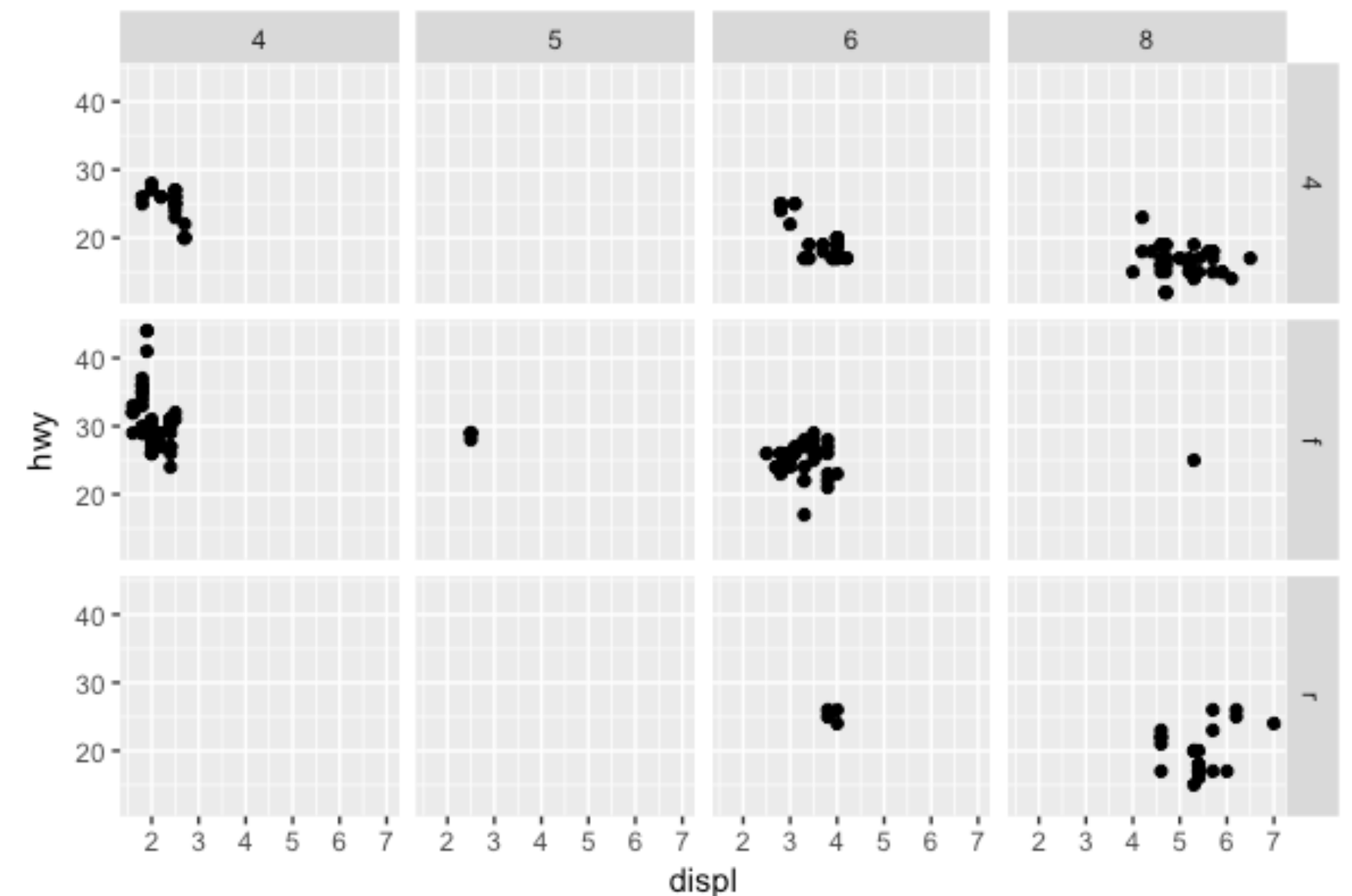
- use **nrow** or **ncol** to specify dimensions
- **?facet_wrap** to see other arguments to control the output



FACETS = SMALL MULTIPLES

- The **facet** functions provide a simple way to create small multiples
 - **facet_wrap**: primarily used to create small multiples based on a single variable
 - **facet_grid**: primarily used to create a small multiples grid based on two variables

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

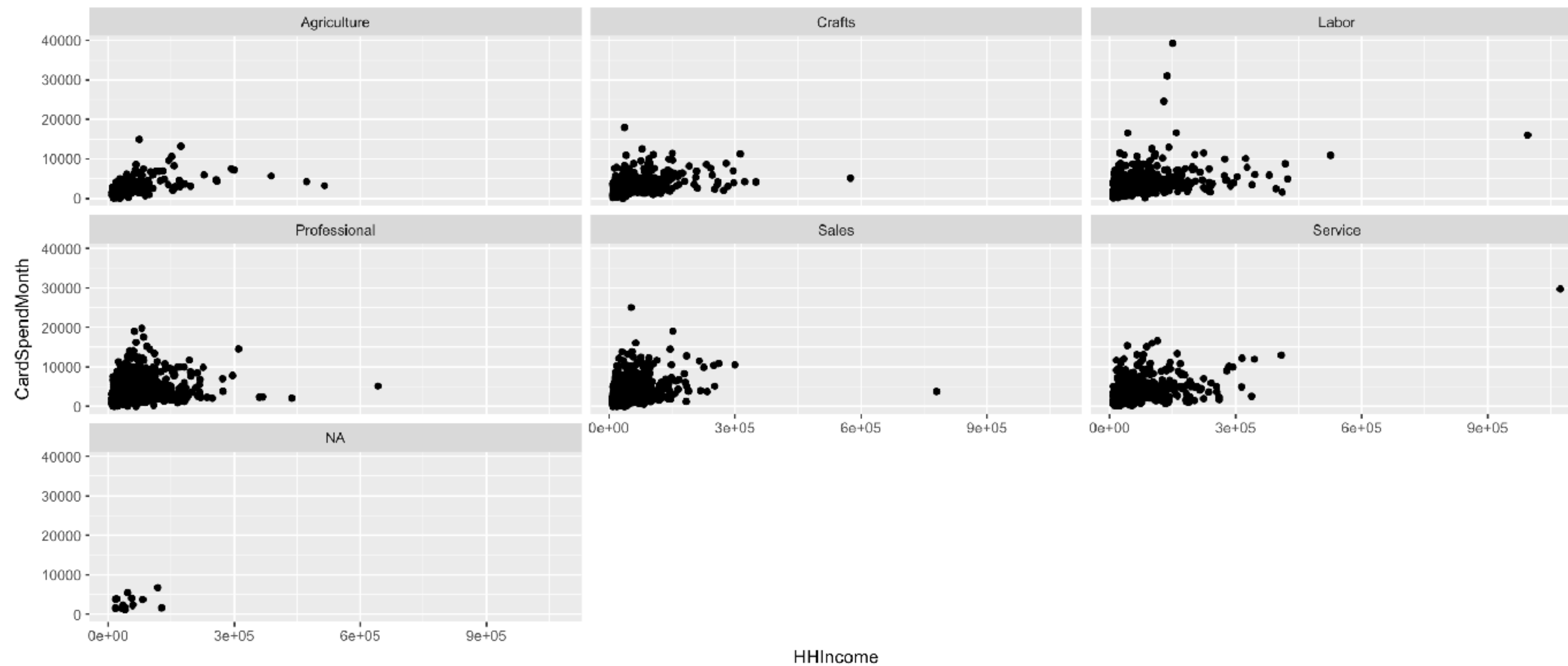


YOUR TURN!

1. Create a scatter plot of *HHIncome* vs *CardSpendMonth* faceted by *JobCategory*.
2. Create a scatter plot of *HHIncome* vs *CardSpendMonth* faceted by *JobCategory* and *Gender*.
3. Assess *UnionMember* across each *JobCategory*.

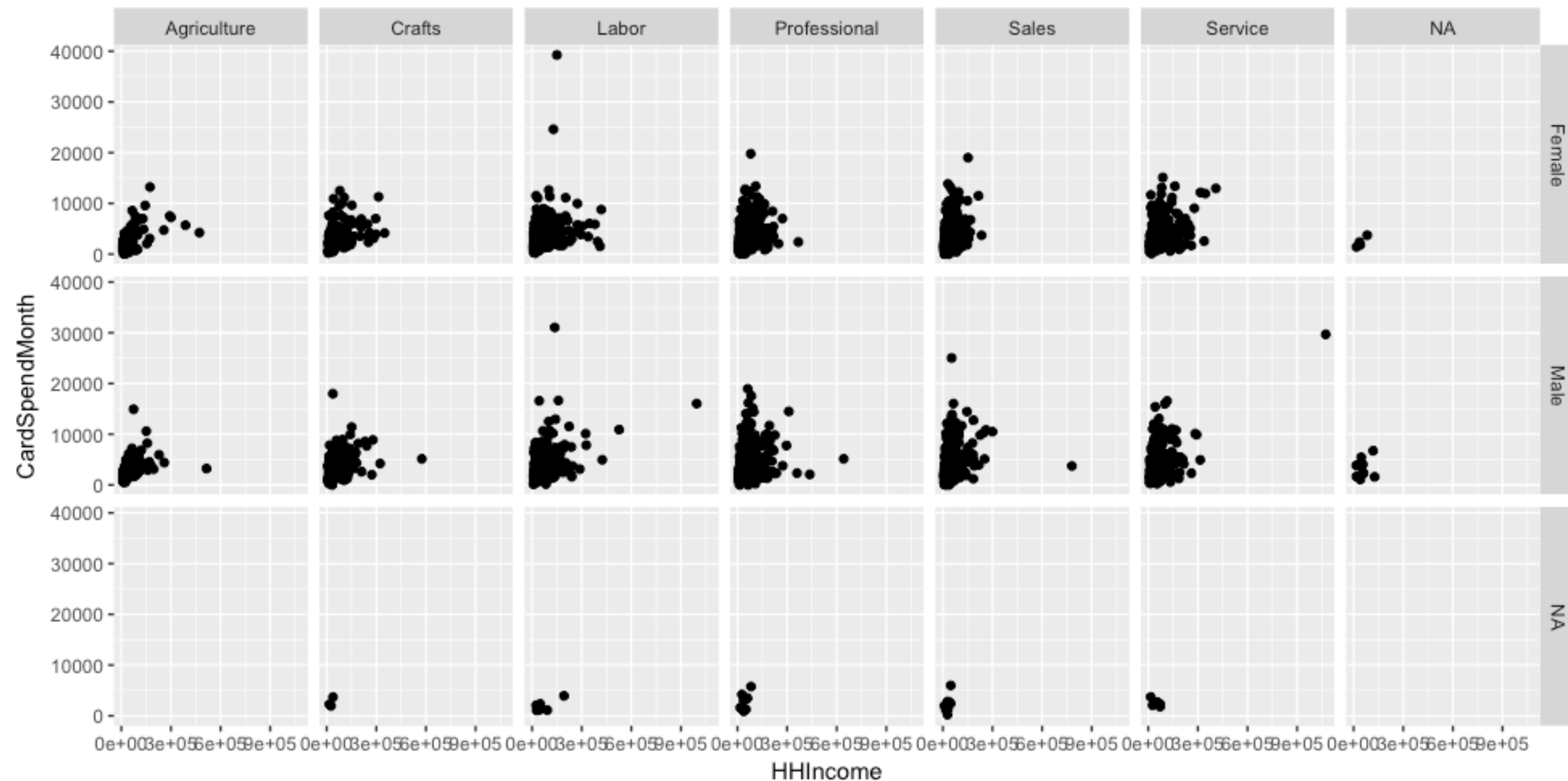
SOLUTION

```
# 2. Create a scatter plot of HHIncome vs CardSpendMonth faceted by JobCategory & Gender.  
ggplot(customer, aes(x = HHIncome, y = CardSpendMonth)) +  
  geom_point() +  
  facet_wrap(~ JobCategory)
```



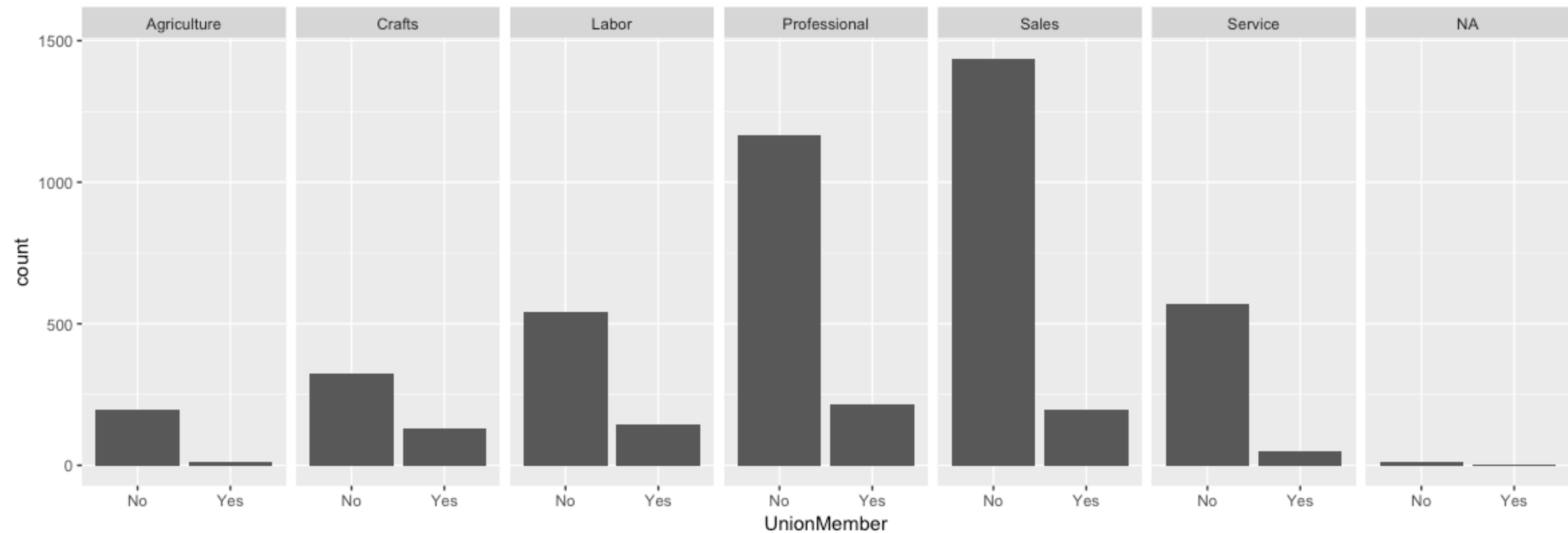
SOLUTION

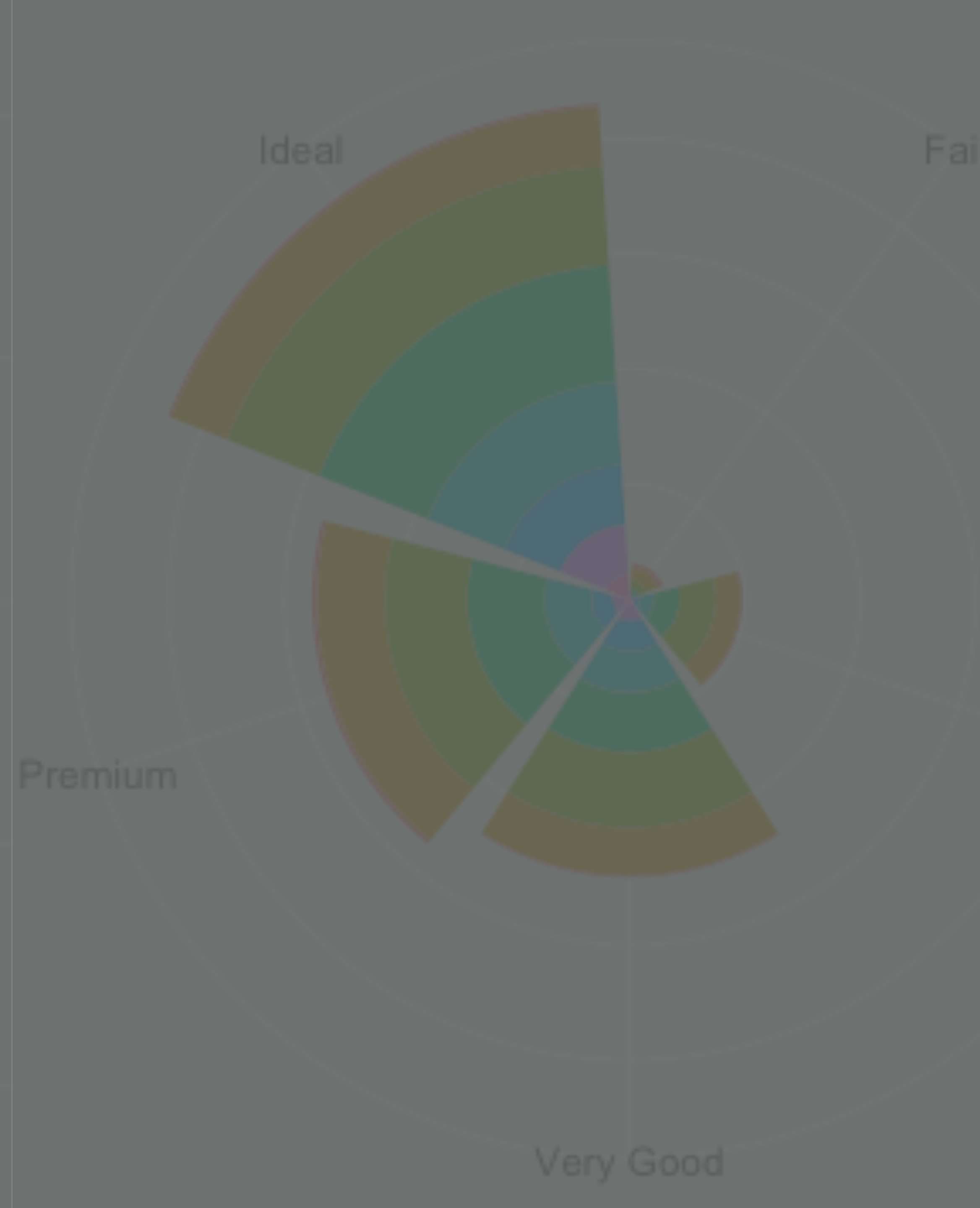
```
# 2. Create a scatter plot of HHIncome vs CardSpendMonth faceted by JobCategory & Gender.  
ggplot(customer, aes(x = HHIncome, y = CardSpendMonth)) +  
  geom_point() +  
  facet_grid(Gender ~ JobCategory)
```



SOLUTION

```
# 3. Assess UnionMember across each JobCategory  
ggplot(customer, aes(x = UnionMember)) +  
  geom_bar() +  
  facet_wrap(~ JobCategory)
```

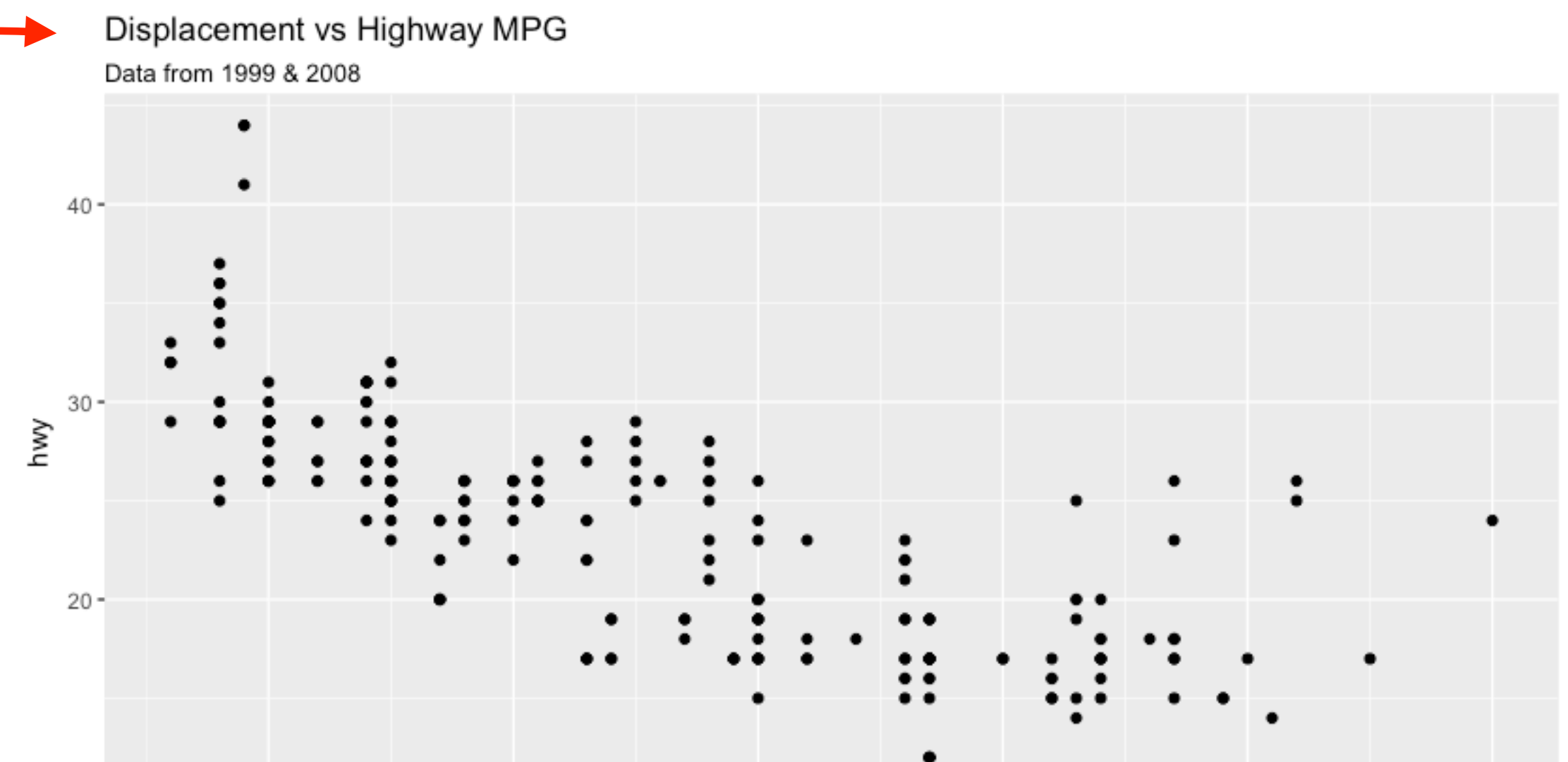
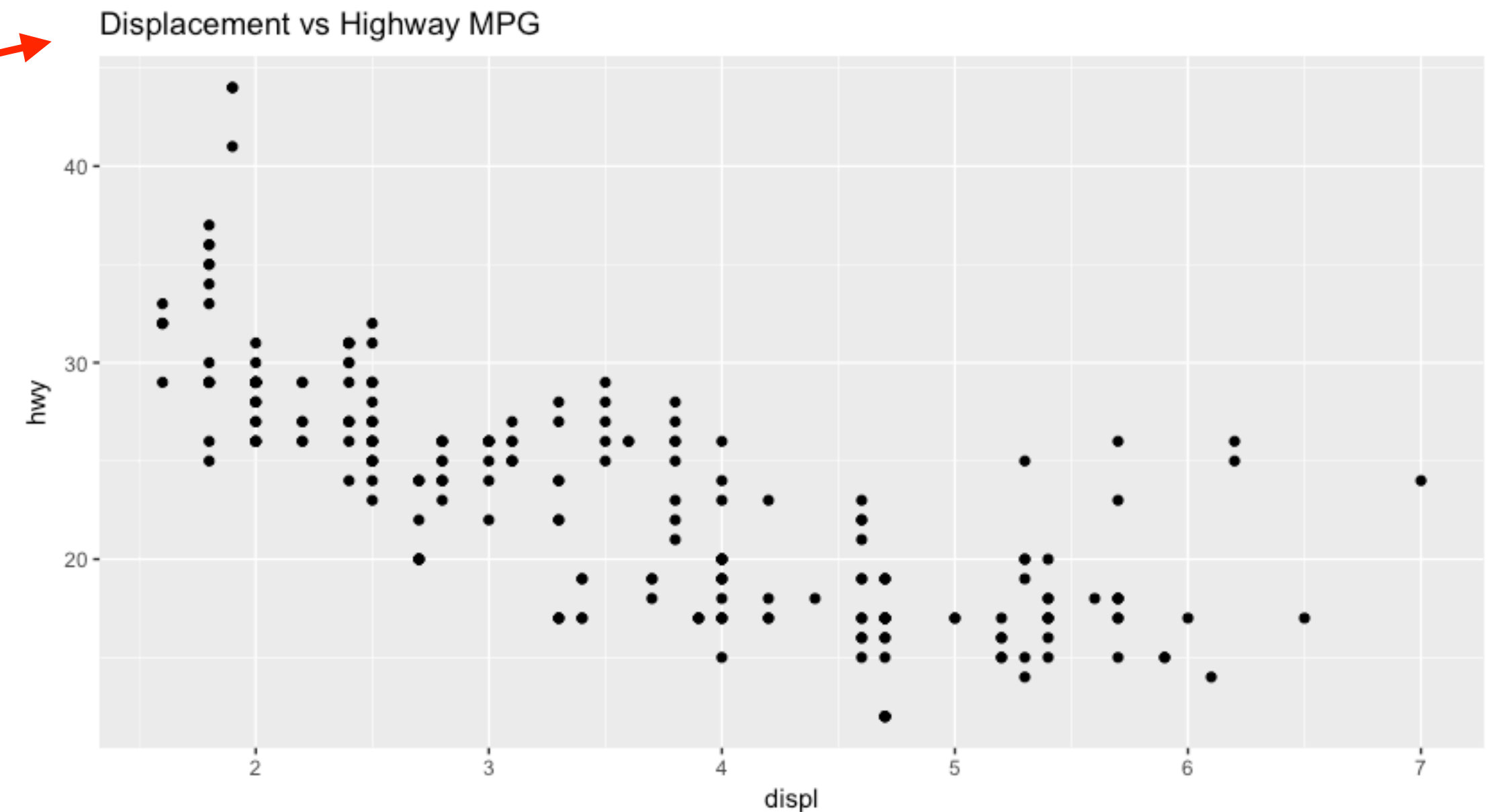




ADDING TITLES

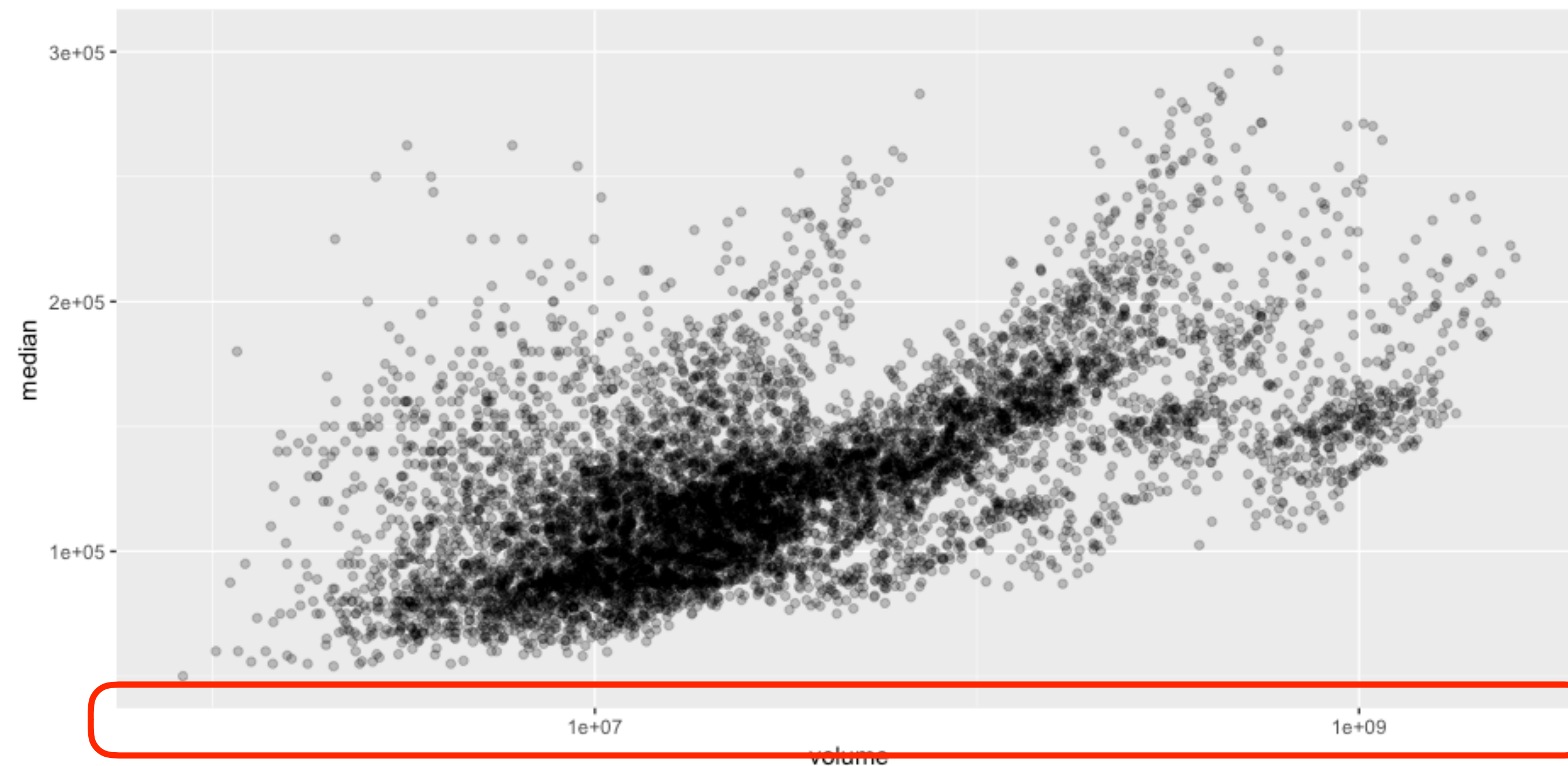
```
# top  
ggplot(data = mpg, aes(x = class, y = hwy)) +  
  geom_point() +  
  ggtitle("Displacement vs Highway MPG")
```

```
# bottom  
ggplot(data = mpg, aes(x = class, y = hwy)) +  
  geom_boxplot() +  
  ggtitle("Displacement vs Highway MPG",  
    subtitle = "Data from 1999 & 2008")
```



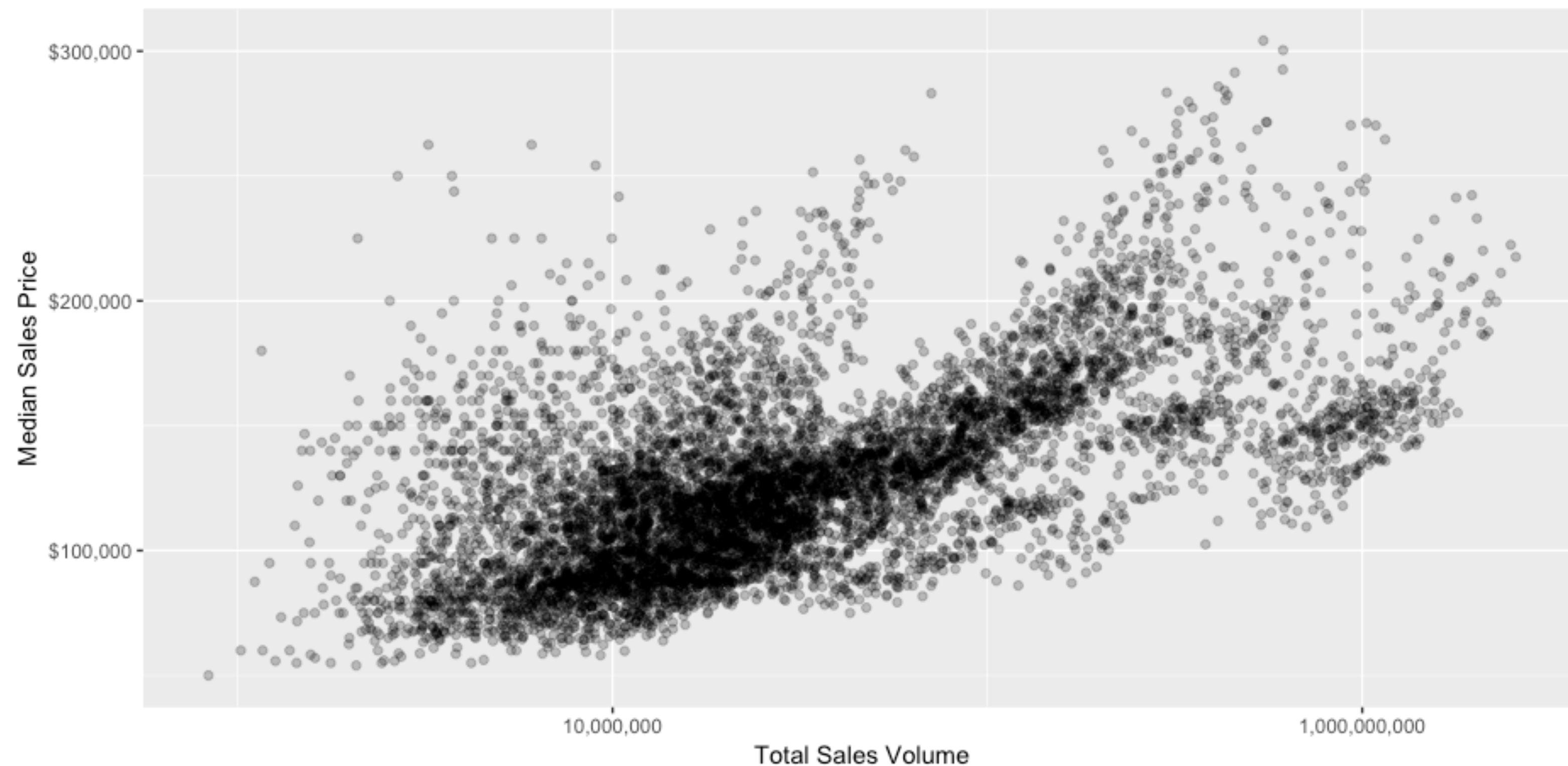
ADJUSTING AXIS SCALES

```
ggplot(data = txhousing, aes(x = volume, y = median)) +  
  geom_point(alpha = .25) +  
  scale_x_log10()
```



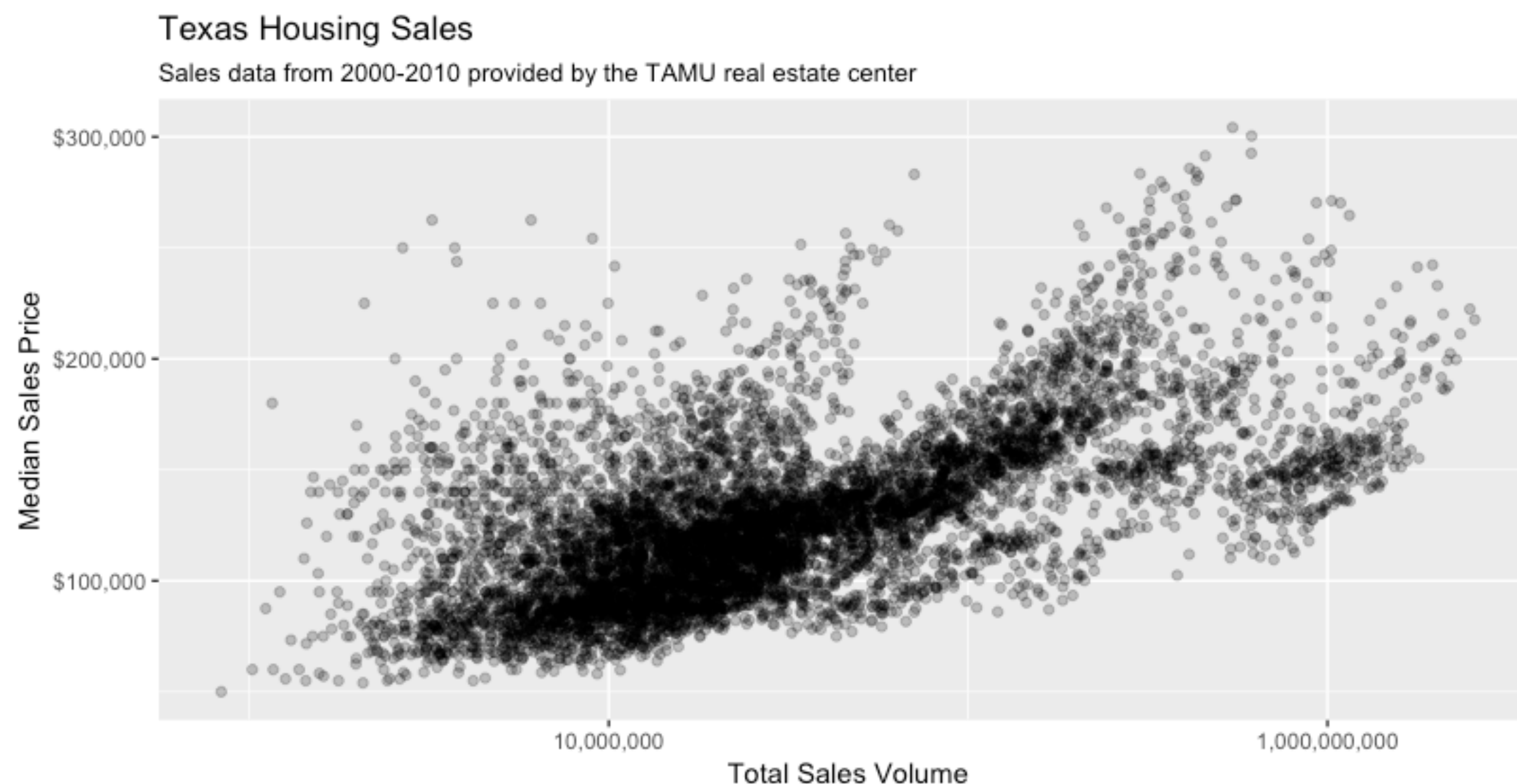
ADJUSTING AXIS TITLES & LABELS

```
ggplot(data = txhousing, aes(x = volume, y = median)) +  
  geom_point(alpha = .25) +  
  scale_y_continuous(name = "Median Sales Price", labels = scales::dollar) +  
  scale_x_log10(name = "Total Sales Volume", labels = scales::comma)
```



PUT IT ALL TOGETHER

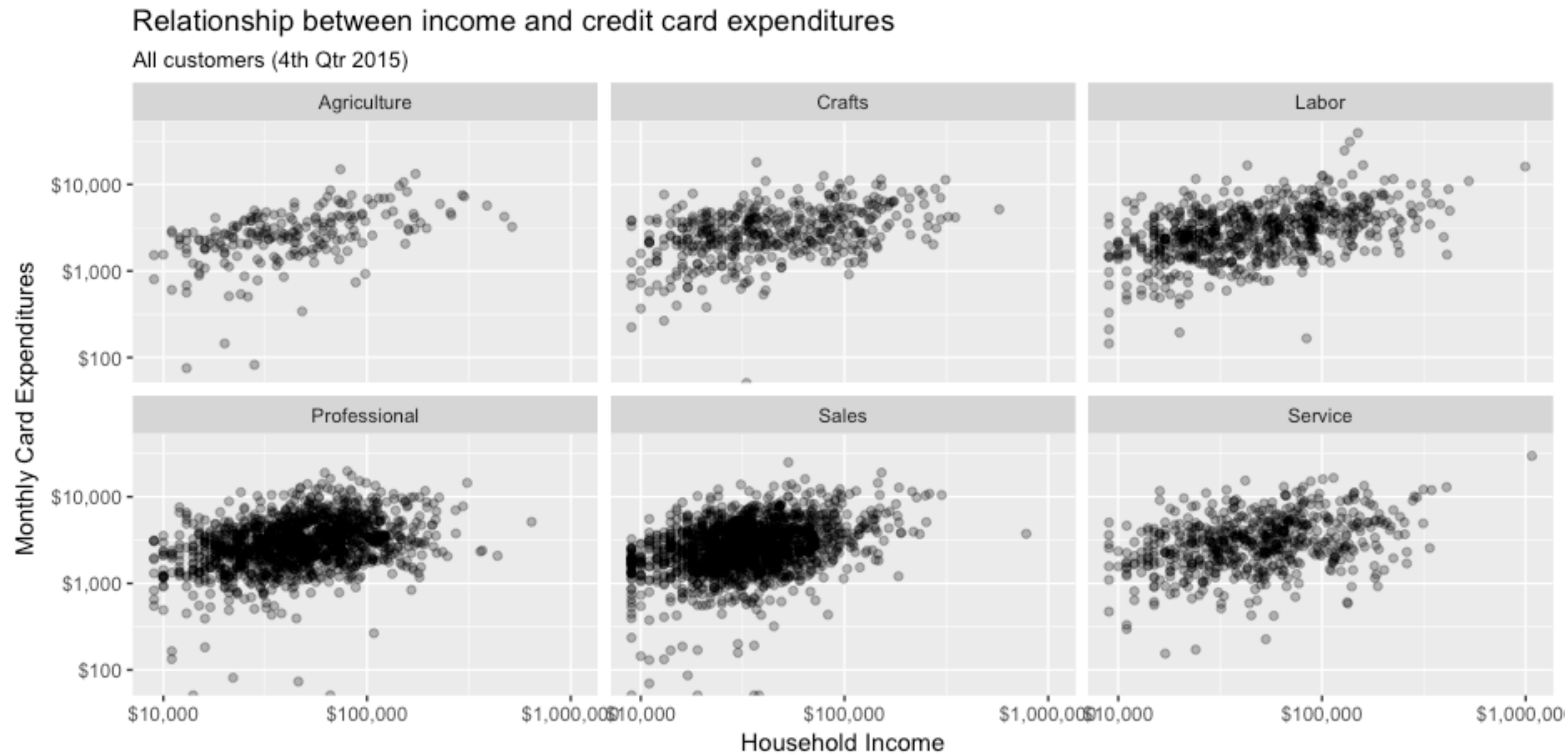
```
ggplot(data = txhousing, aes(x = volume, y = median)) +  
  geom_point(alpha = .25) +  
  scale_y_continuous(name = "Median Sales Price", labels = scales::dollar) +  
  scale_x_log10(name = "Total Sales Volume", labels = scales::comma) +  
  ggtitle("Texas Housing Sales",  
    subtitle = "Sales data from 2000-2010 provided by the TAMU real estate center")
```



YOUR TURN!

1. Remove all missing values from the customer data and then...
2. Create a scatter plot of *HHIncome* vs *CardSpendMonth* faceted by *JobCategory* and...
3. add a title, subtitle, and nicely format the axes.

```
customer %>%
  na.omit() %>%
  ggplot(aes(x = HHIncome, y = CardSpendMonth)) +
  geom_point(alpha = .3) +
  facet_wrap(~ JobCategory) +
  scale_x_log10("Household Income", labels = scales::dollar) +
  scale_y_log10("Monthly Card Expenditures", labels = scales::dollar) +
  ggtitle("Relationship between income and credit card expenditures",
    subtitle = "All customers (4th Qtr 2015)")
```

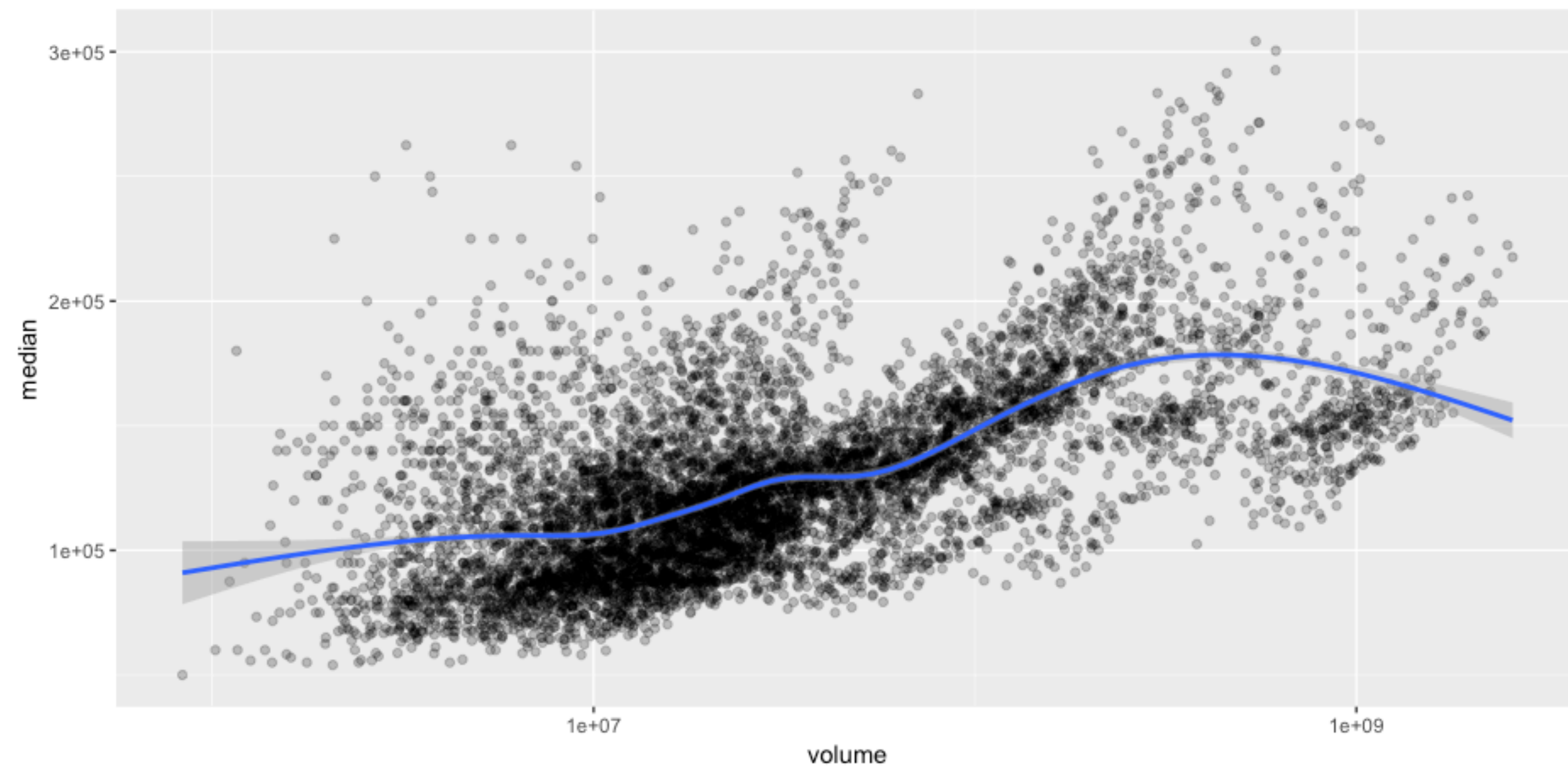




OVERPLOTING

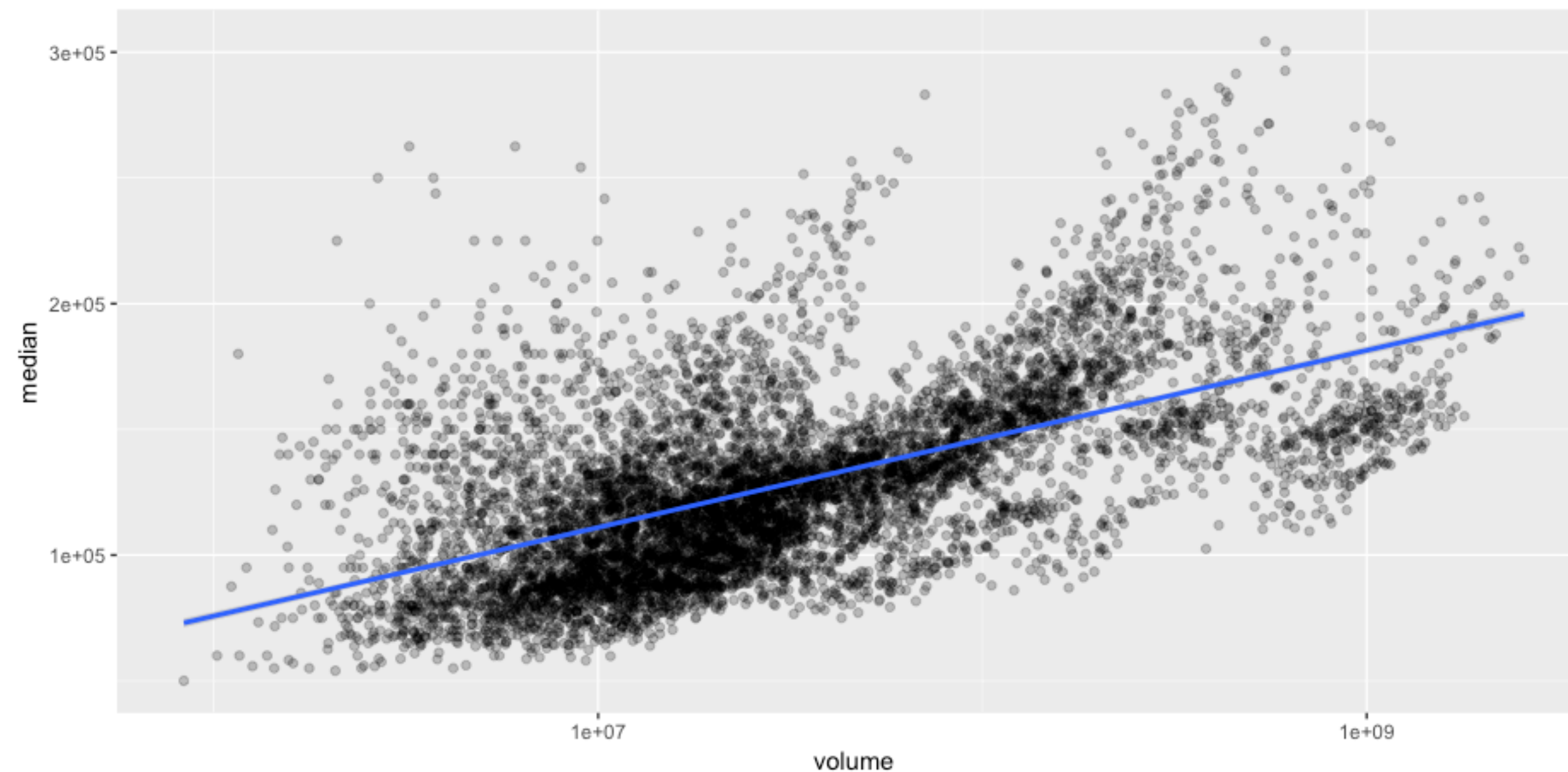
LAYERING HELPS DISPLAY PATTERNS

```
ggplot(data = txhousing, aes(x = volume, y = median)) +  
  geom_point(alpha = .25) +  
  scale_x_log10() +  
  geom_smooth()
```



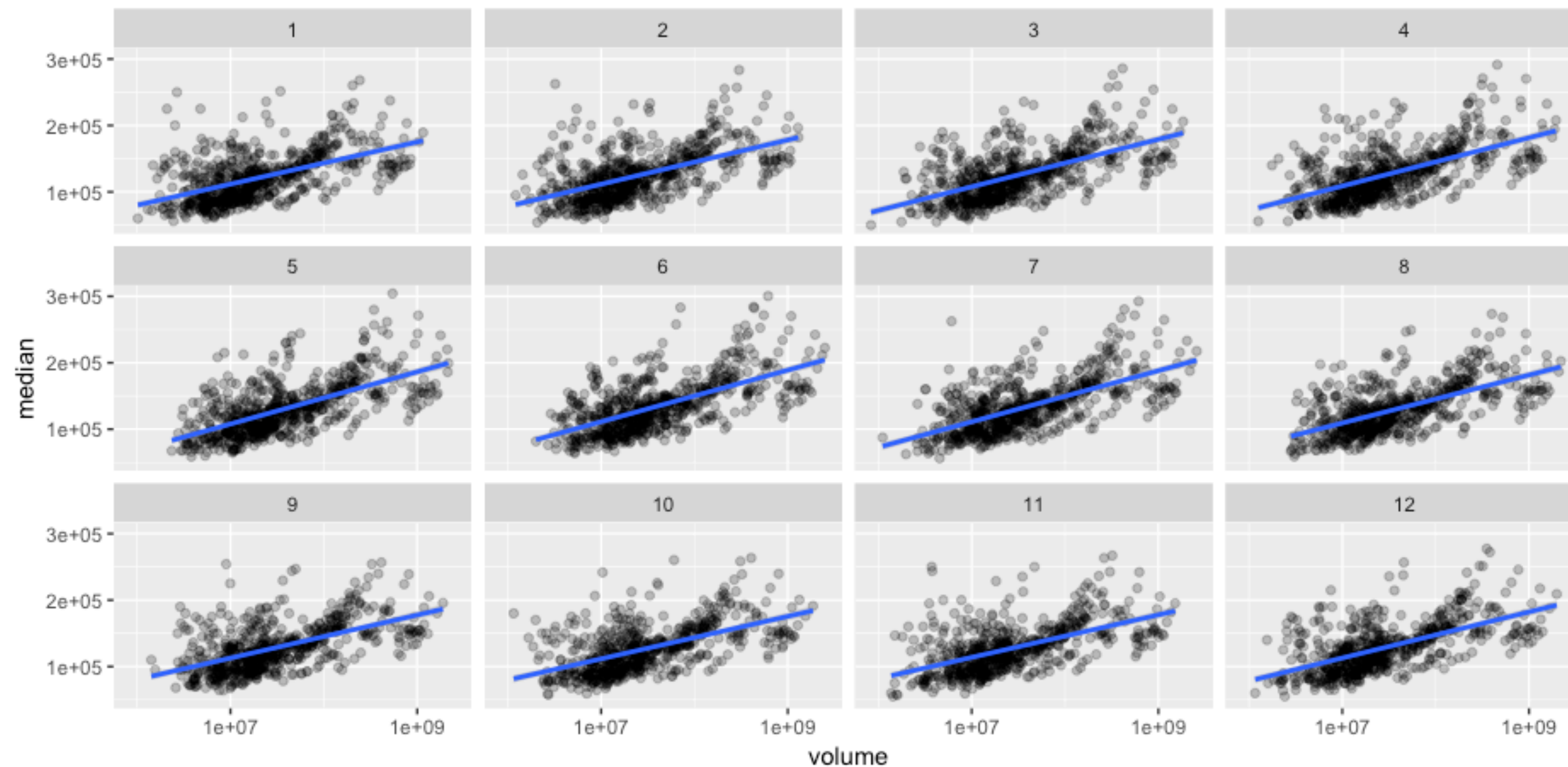
LAYERING HELPS DISPLAY PATTERNS

```
ggplot(data = txhousing, aes(x = volume, y = median)) +  
  geom_point(alpha = .25) +  
  scale_x_log10() +  
  geom_smooth(method = "lm")
```



LAYERING HELPS DISPLAY PATTERNS

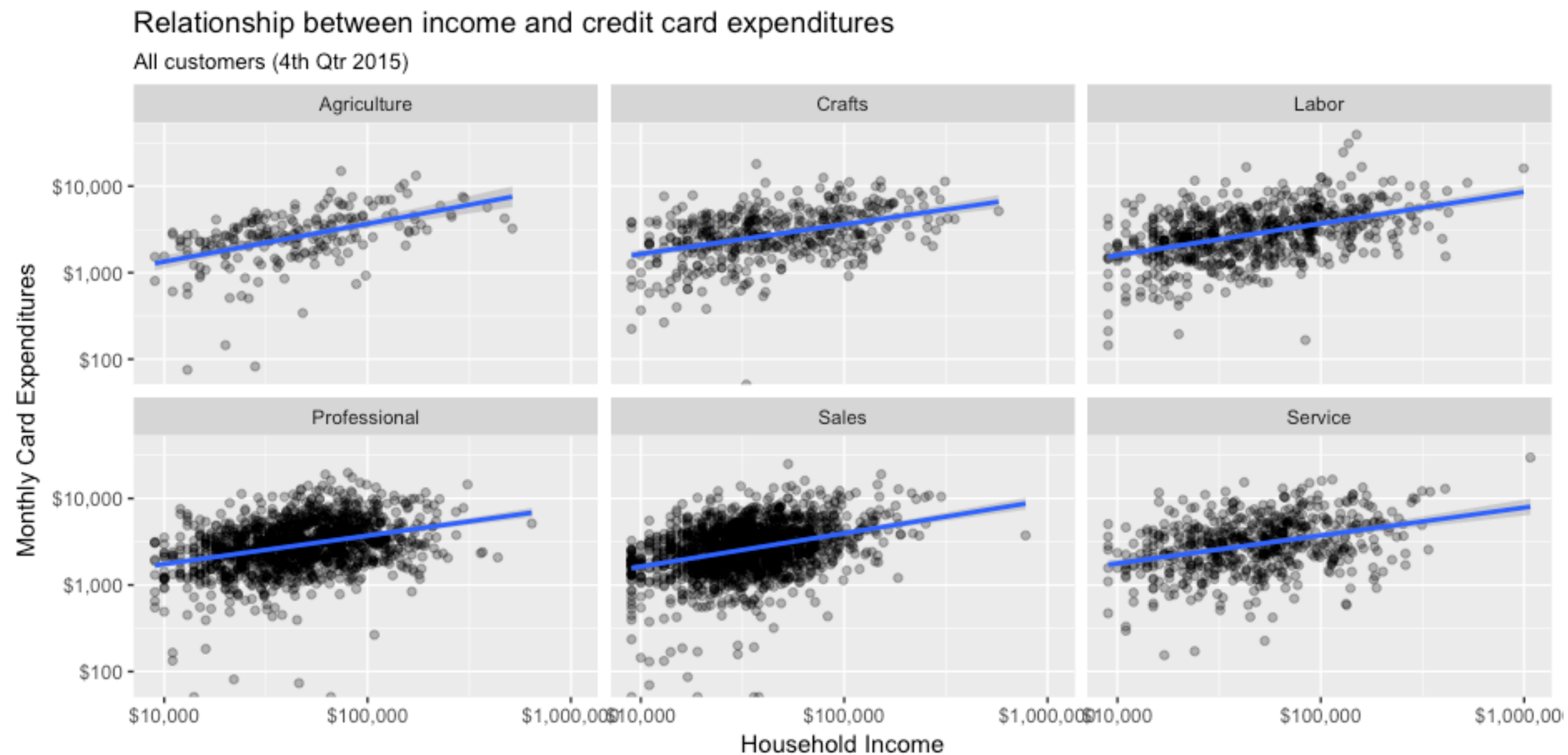
```
ggplot(data = txhousing, aes(x = volume, y = median)) +  
  geom_point(alpha = .25) +  
  scale_x_log10() +  
  geom_smooth(method = "lm") +  
  facet_wrap(~ month)
```



YOUR TURN!

1. Remove all missing values from the customer data and then...
2. Create a scatter plot of *HHIncome* vs *CardSpendMonth* faceted by *JobCategory* and...
3. add a title, subtitle, and nicely format the axes and...
4. add a linear line to assess if the slope changes across *JobCategory*


```
customer %>%
  na.omit() %>%
  ggplot(aes(x = HHIncome, y = CardSpendMonth)) +
  geom_point(alpha = .3) +
  geom_smooth(method = "lm") +
  facet_wrap(~ JobCategory) +
  scale_x_log10("Household Income", labels = scales::dollar) +
  scale_y_log10("Monthly Card Expenditures", labels = scales::dollar) +
  ggtitle("Relationship between income and credit card expenditures",
    subtitle = "All customers (4th Qtr 2015)")
```




LEVERAGE HELP AS YOU'RE LEARNING

Help >> Cheatsheets >> Data Visualization with ggplot2

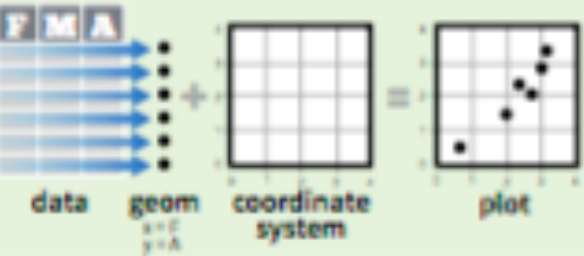
Data Visualization with ggplot2

Cheat Sheet

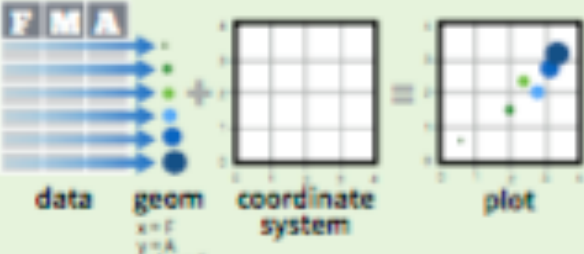


Basics

ggplot2 is based on the **grammar of graphics**, the idea that you can build every graph from the same few components: a **data** set, a set of **geoms**—visual marks that represent data points, and a **coordinate system**.



To display data values, map variables in the data set to aesthetic properties of the geom like **size**, **color**, and **x** and **y** locations.



Build a graph with **ggplot()** or **qplot()**

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

Geoms

Use a geom to represent data points, use the geom's aesthetic properties to represent variables. Each function returns a layer.

Graphical Primitives	Two Variables
a + geom_blank() (Useful for expanding limits)	Continuous X, Continuous Y e <- ggplot(mpg, aes(cty, hwy))
a + geom_curve(aes(yend = lat + delta_lat, xend = long + delta_long, curvature = z)) x, xend, y, yend, alpha, angle, color, curvature, linetype, size	Continuous Bivariate Distribution h <- ggplot(diamonds, aes(carat, price))
b + geom_path(lineend = "butt", linejoin = "round", linemitre = 1) x, y, alpha, color, group, linetype, size	h + geom_bin2d(binwidth = c(0.25, 500)) x, y, alpha, color, fill, linetype, size, weight
b + geom_polygon(aes(group = group)) x, y, alpha, color, fill, group, linetype, size	h + geom_density2d() x, y, alpha, colour, group, linetype, size
a + geom_rect(aes(xmin = long, ymin = lat, xmax = long + delta_long, ymax = lat + delta_lat)) xmax, xmin, ymax, ymin, alpha, color, fill, linetype, size	h + geom_hex() x, y, alpha, colour, fill, size
b + geom_ribbon(aes(ymin = unemploy - 900, ymax = unemploy + 900)) x, ymax, ymin, alpha, color, fill, group, linetype, size	Continuous Function i <- ggplot(economics, aes(date, unemploy))
a + geom_segment(aes(yend = lat + delta_lat, xend = long + delta_long)) x, xend, y, yend, alpha, color, linetype, size	i + geom_area() x, y, alpha, color, fill, linetype, size
a + geom_spoke(aes(yend = lat + delta_lat, xend = long + delta_long)) x, y, angle, radius, alpha, color, linetype, size	i + geom_line() x, y, alpha, color, group, linetype, size
	i + geom_step(direction = "hv") x, y, alpha, color, group, linetype, size
	Visualizing error df <- data.frame(grp = c("A", "B"), fit = 4:5, se = 1:2) j <- ggplot(df, aes(grp, fit, ymin = fit - se, ymax = fit + se))
	Discrete X, Continuous Y f <- ggplot(mpg, aes(class, hwy))
	f + geom_bar(stat = "identity") x, y, alpha, color, fill, linetype, size, weight
	j + geom_crossbar(fatten = 2) x, y, ymax, ymin, alpha, color, fill, group, linetype, size
	i + geom_errorbar()

WHAT TO REMEMBER



FUNCTIONS TO REMEMBER

Operator/Function	Description
<code>ggplot()</code>	Initializes a ggplot object (creates the blank canvas)
<code>aes()</code>	Creates aesthetic mappings
<code>geom_xx</code>	Geometric shapes to plot the data
<code>color, shape, size, alpha, etc</code>	Aesthetic parameters
<code>facet_wrap, facet_grid</code>	Create small multiples
<code>position</code>	Position argument (primarily used with bar charts)
<code>coord_xx</code>	Functions to adjust the coordinate system
<code>scale_xx</code>	Functions to adjust x and y axis