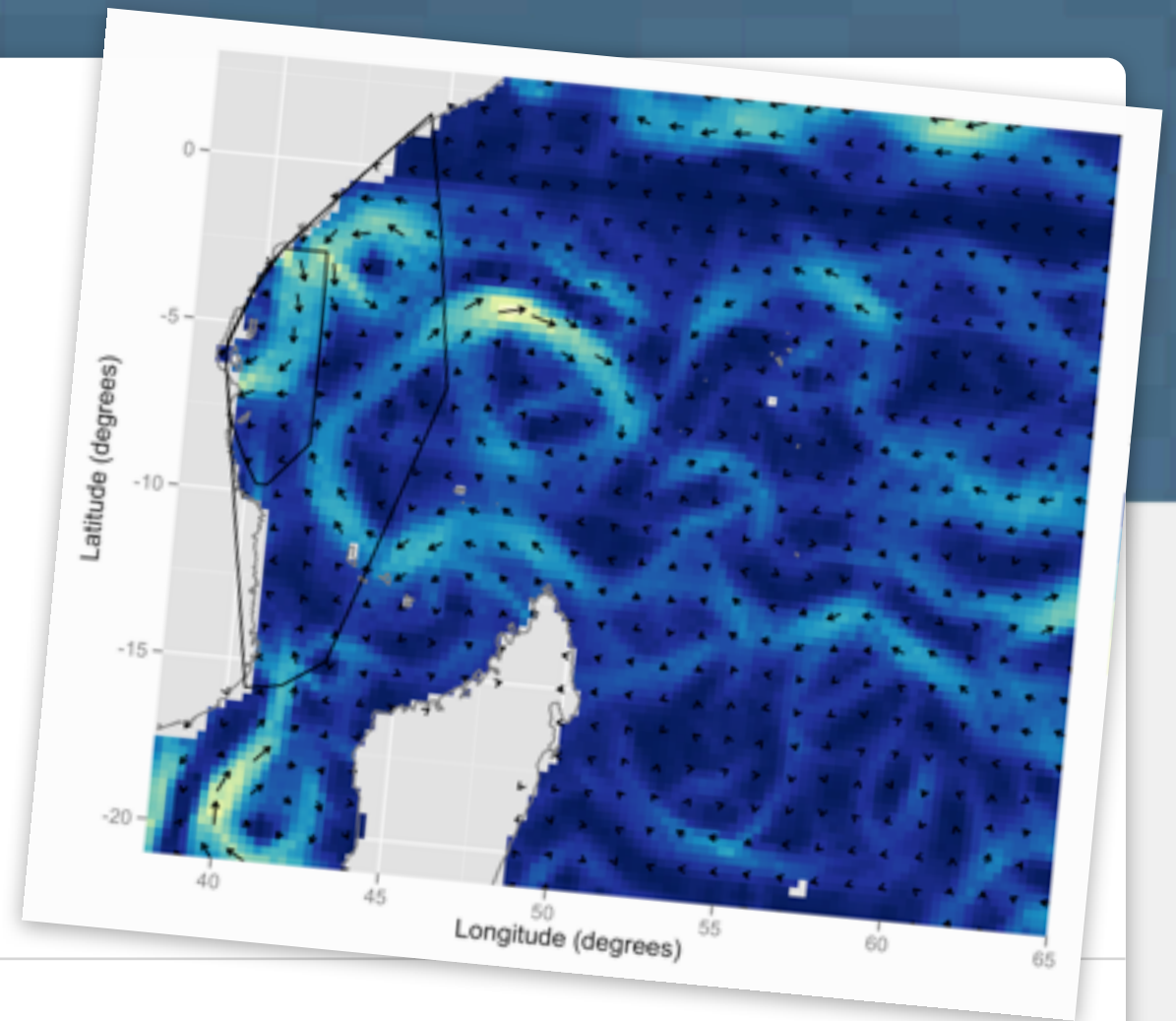


# Visualizing Data

Discover the unexpected in  
your data with ggplot2



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**July 2013**

# 1. Scatterplots

a. Aesthetics, Facets, Geoms

# 2. Histograms and barcharts

a. Parameters, Position adjustments

# 3. Grammar of graphics

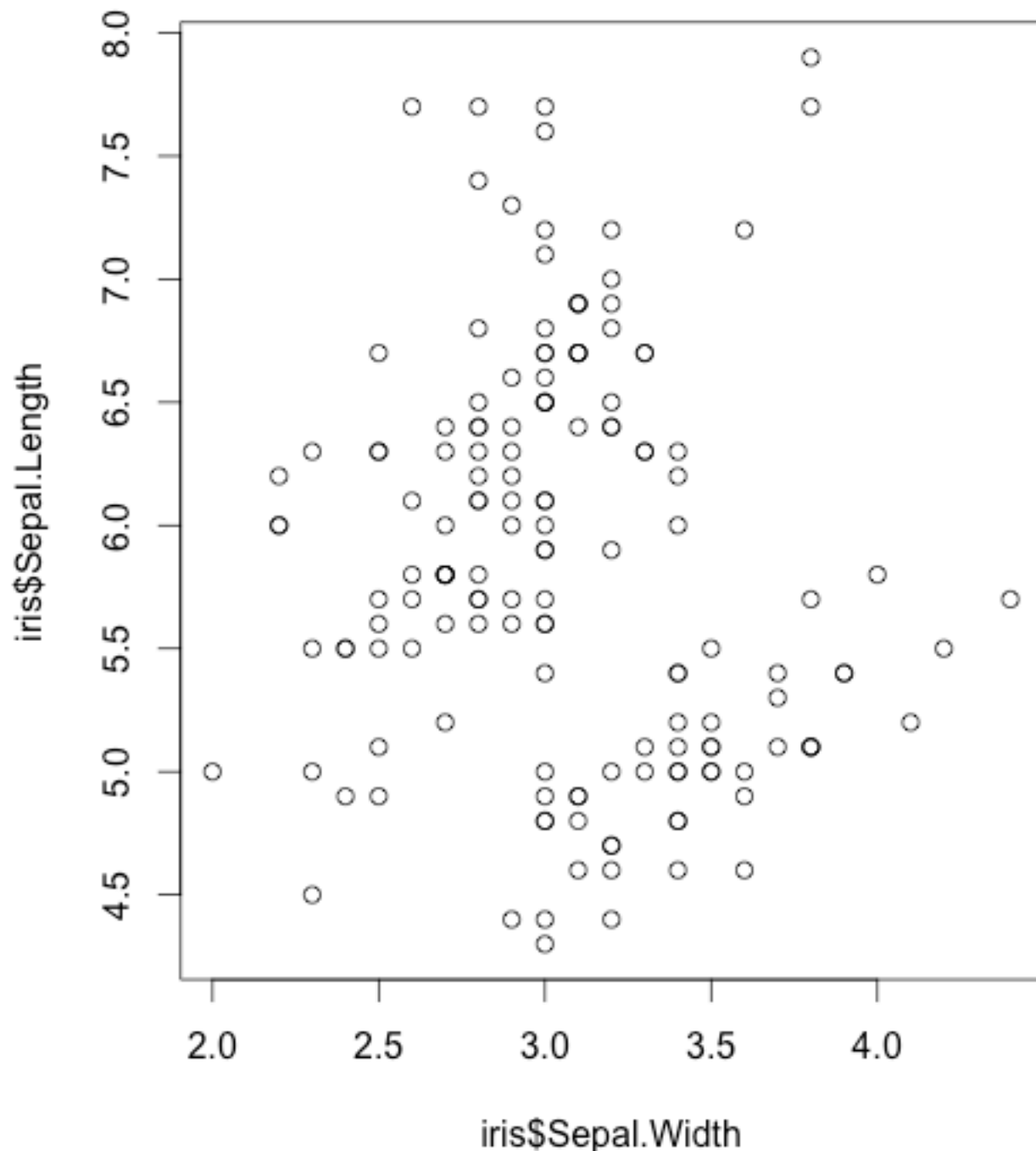
# 4. Layers

# 5. Customizing graphics

The simple graph has brought more information to the data analyst's mind than any other device.

— John Tukey

# plot

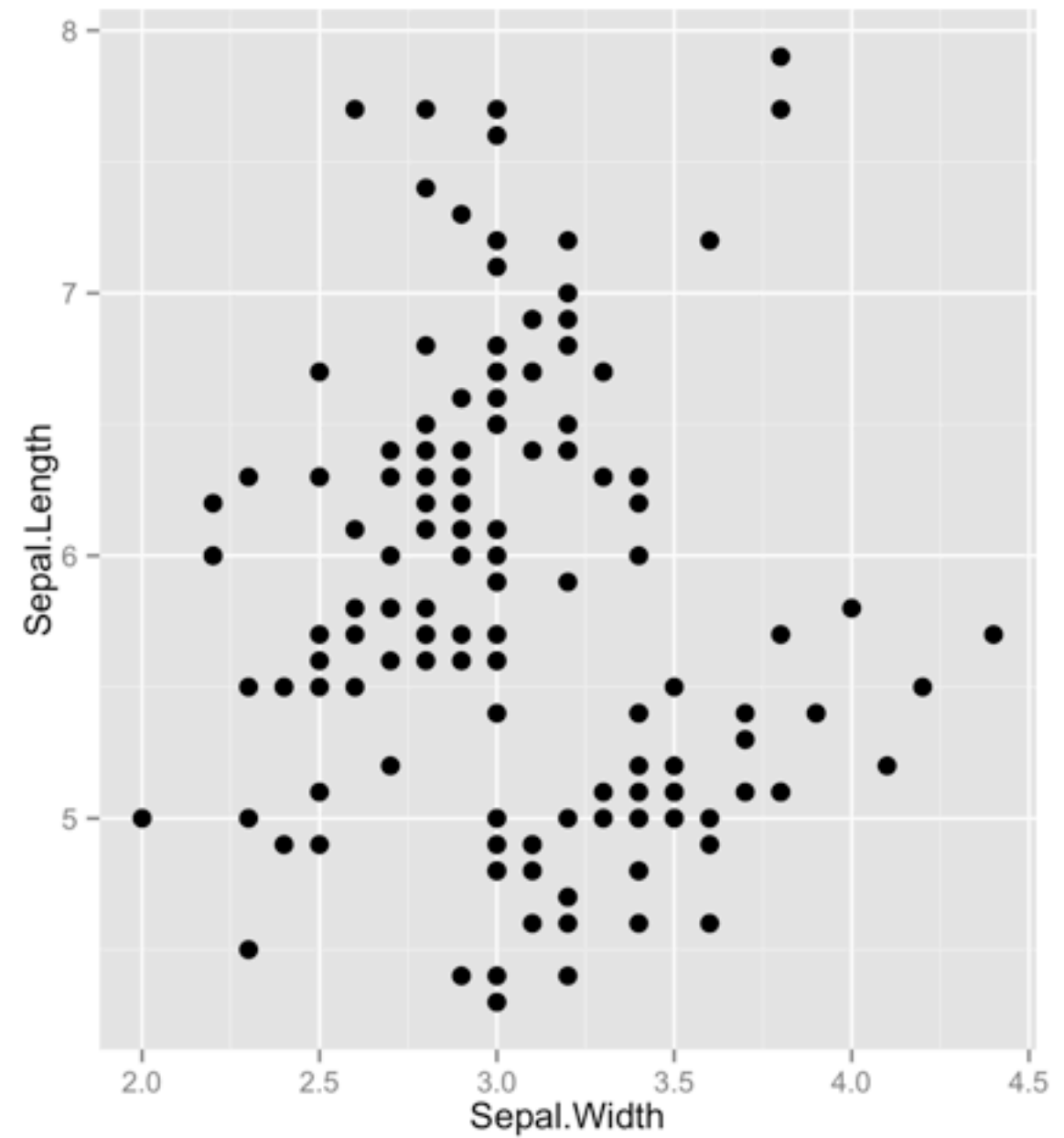
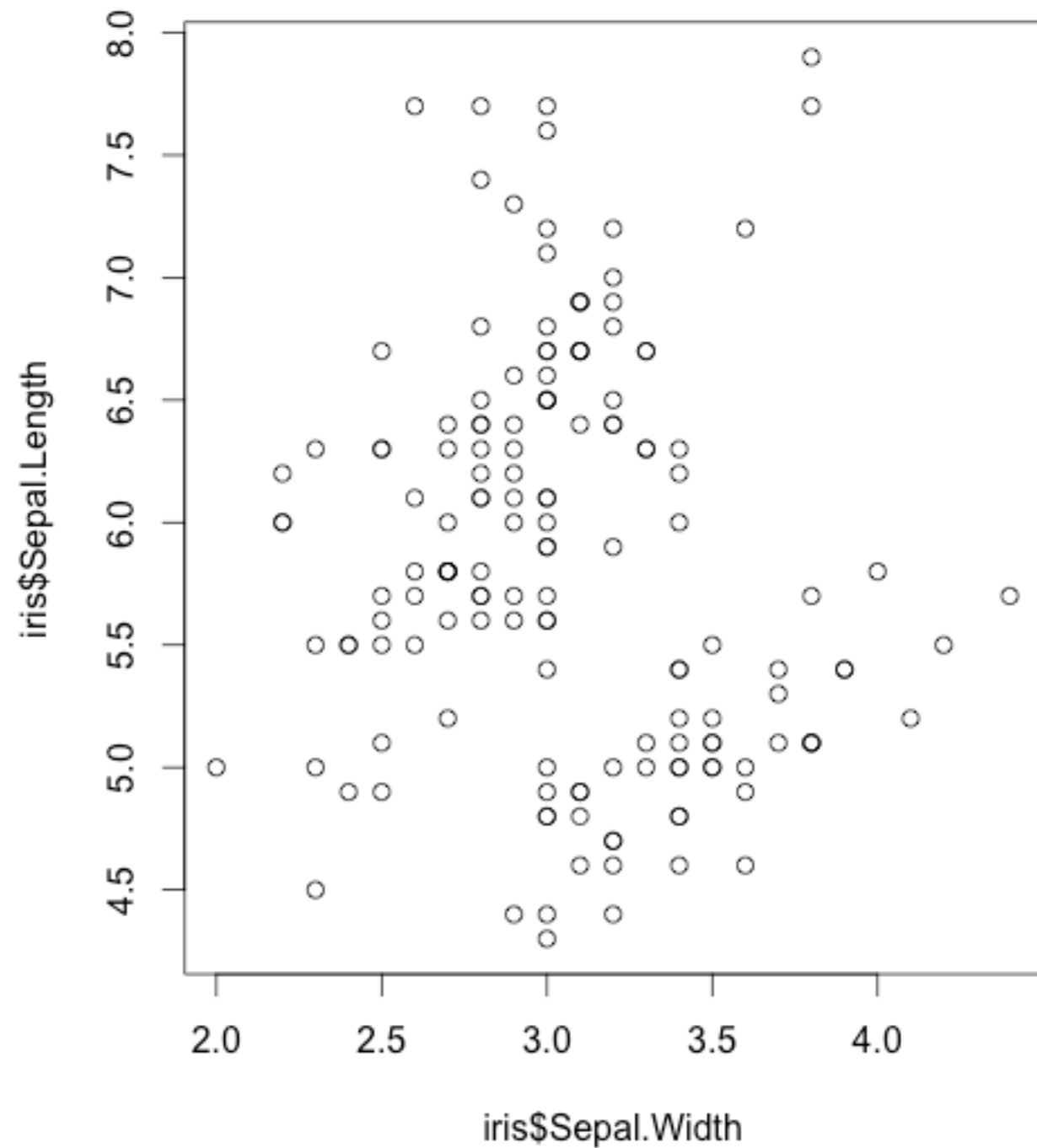


```
plot(iris$Sepal.Width,  
      iris$Sepal.Length)
```

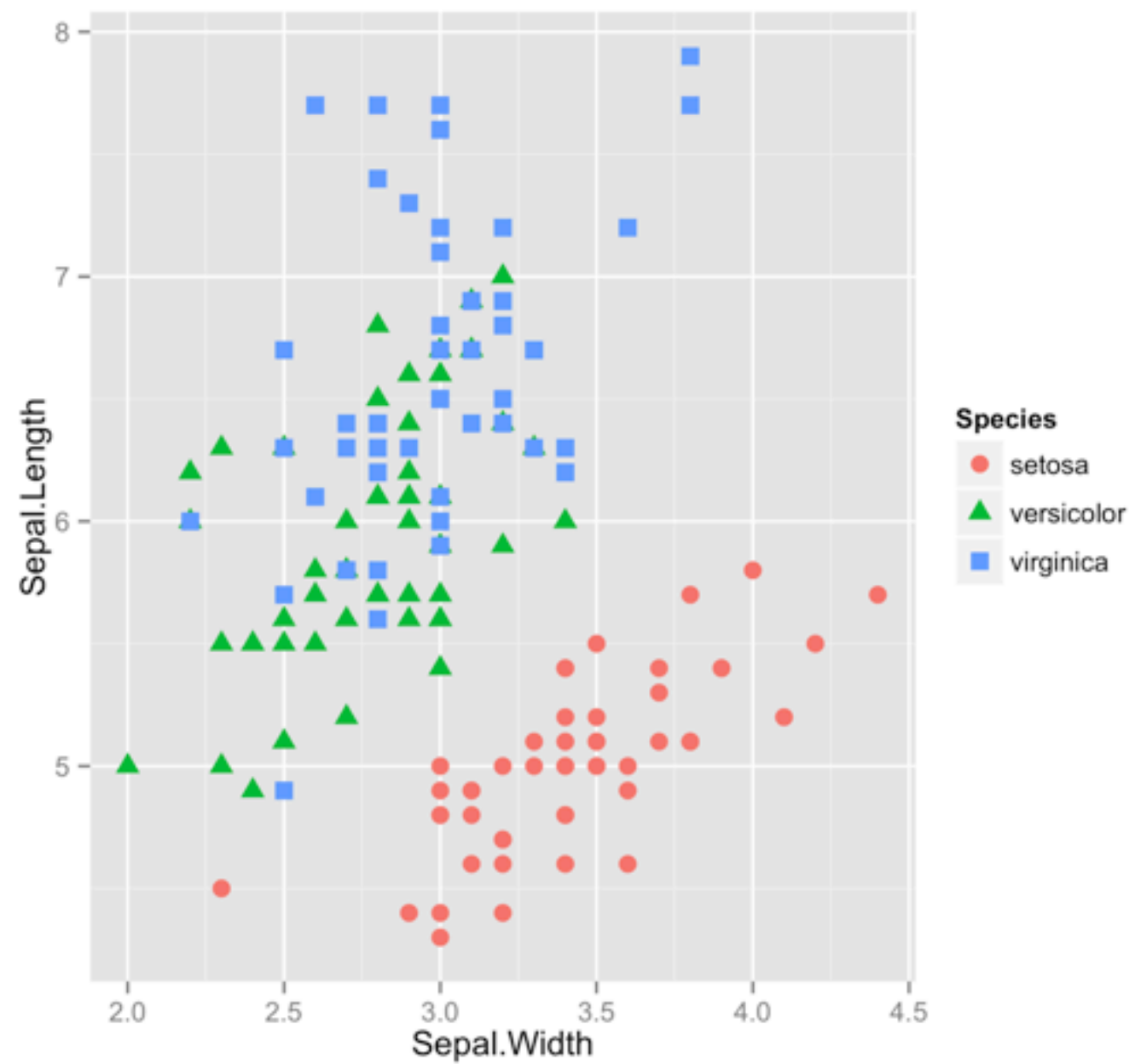
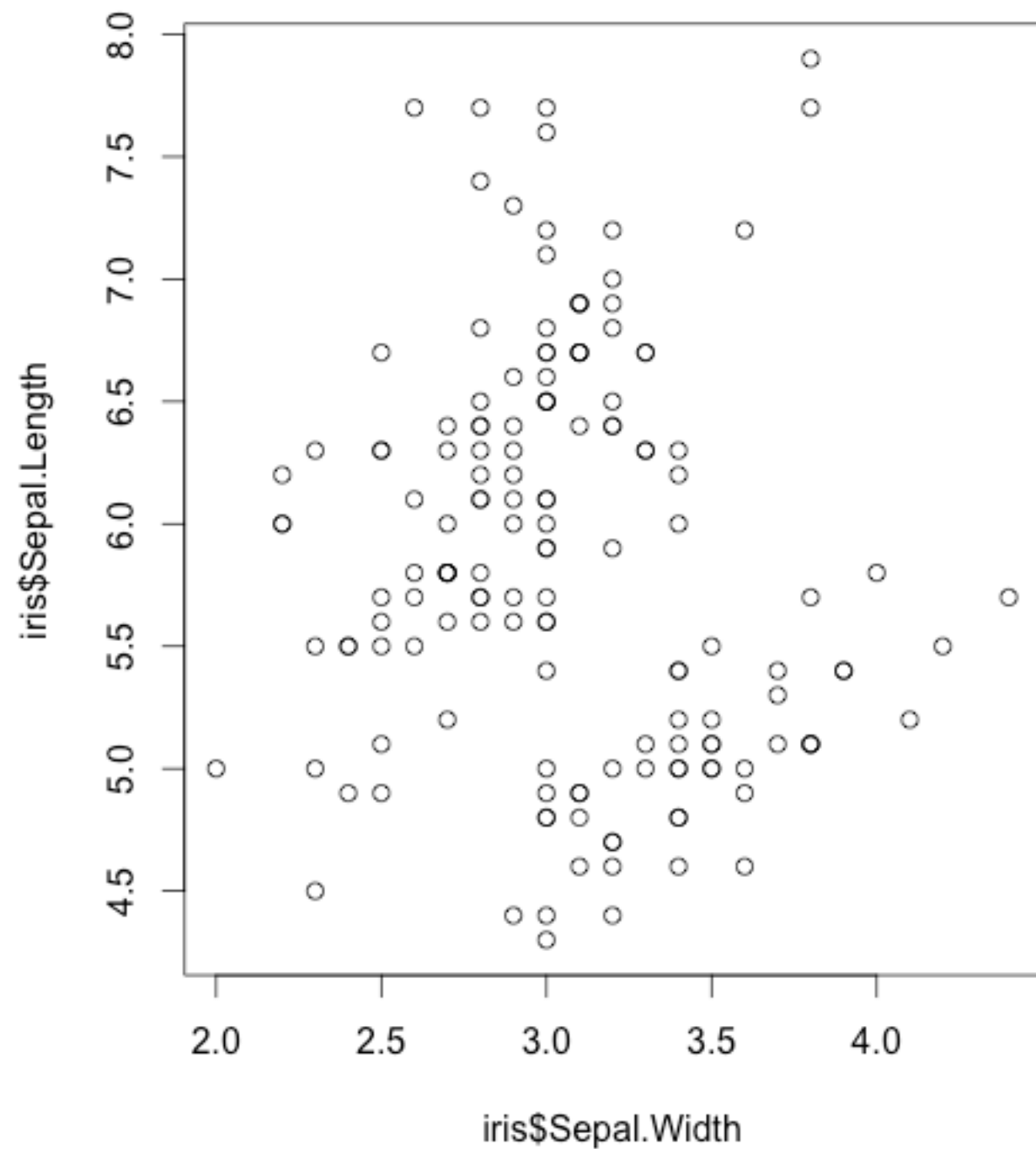
- R's basic plot method
- simple
- does different things in different contexts (usually in a helpful way)
- difficult to customize



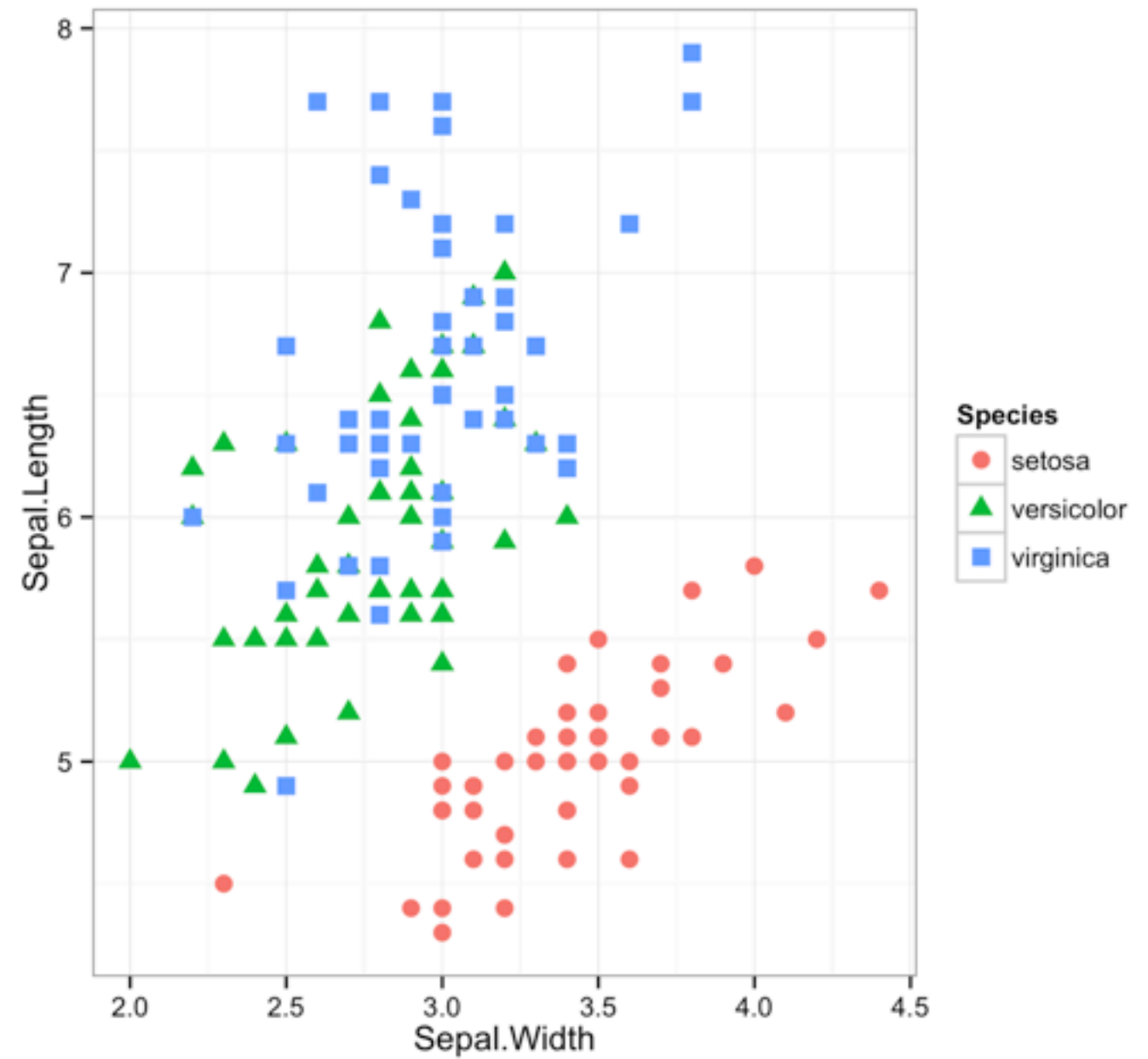
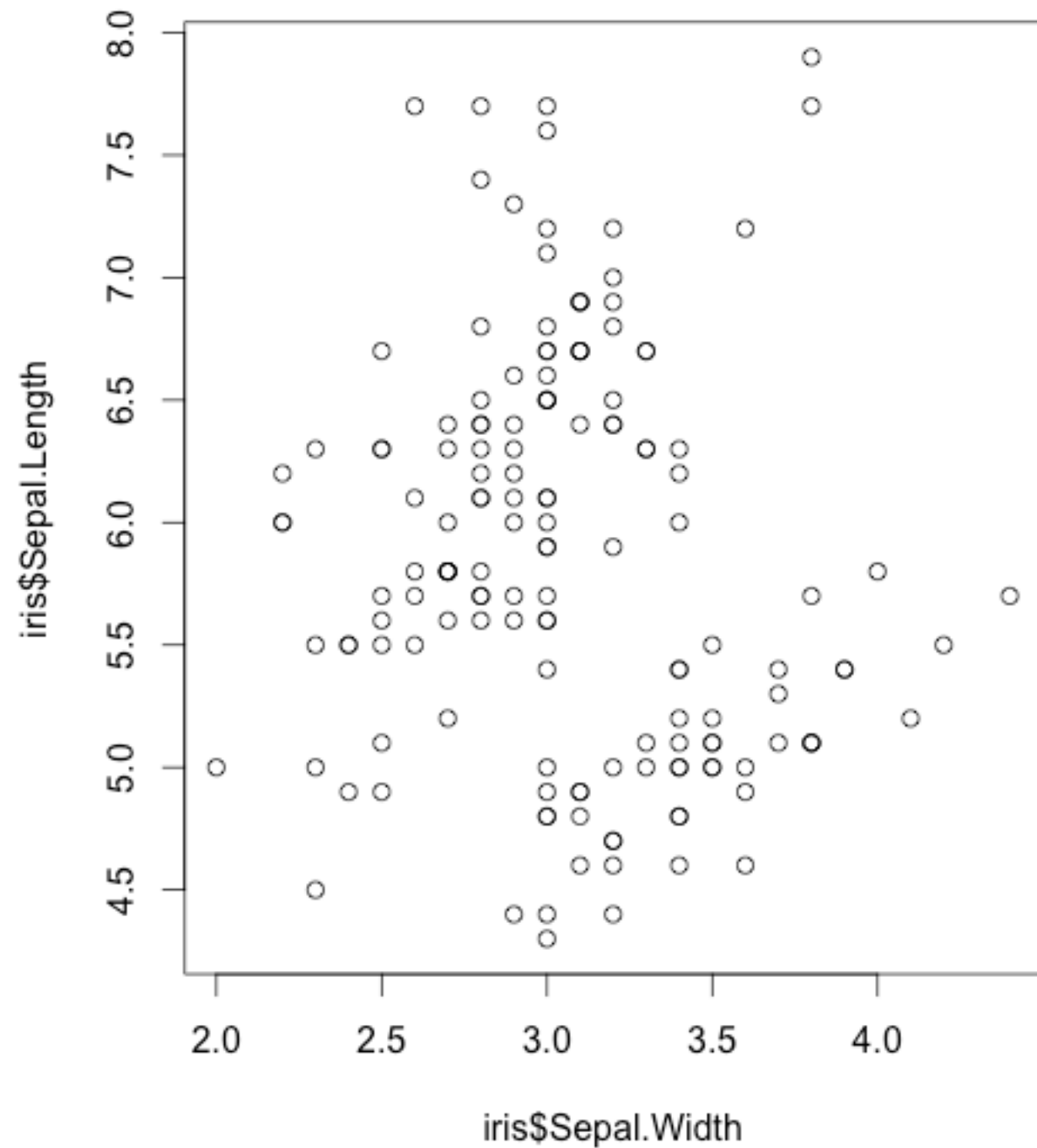
# ggplot2



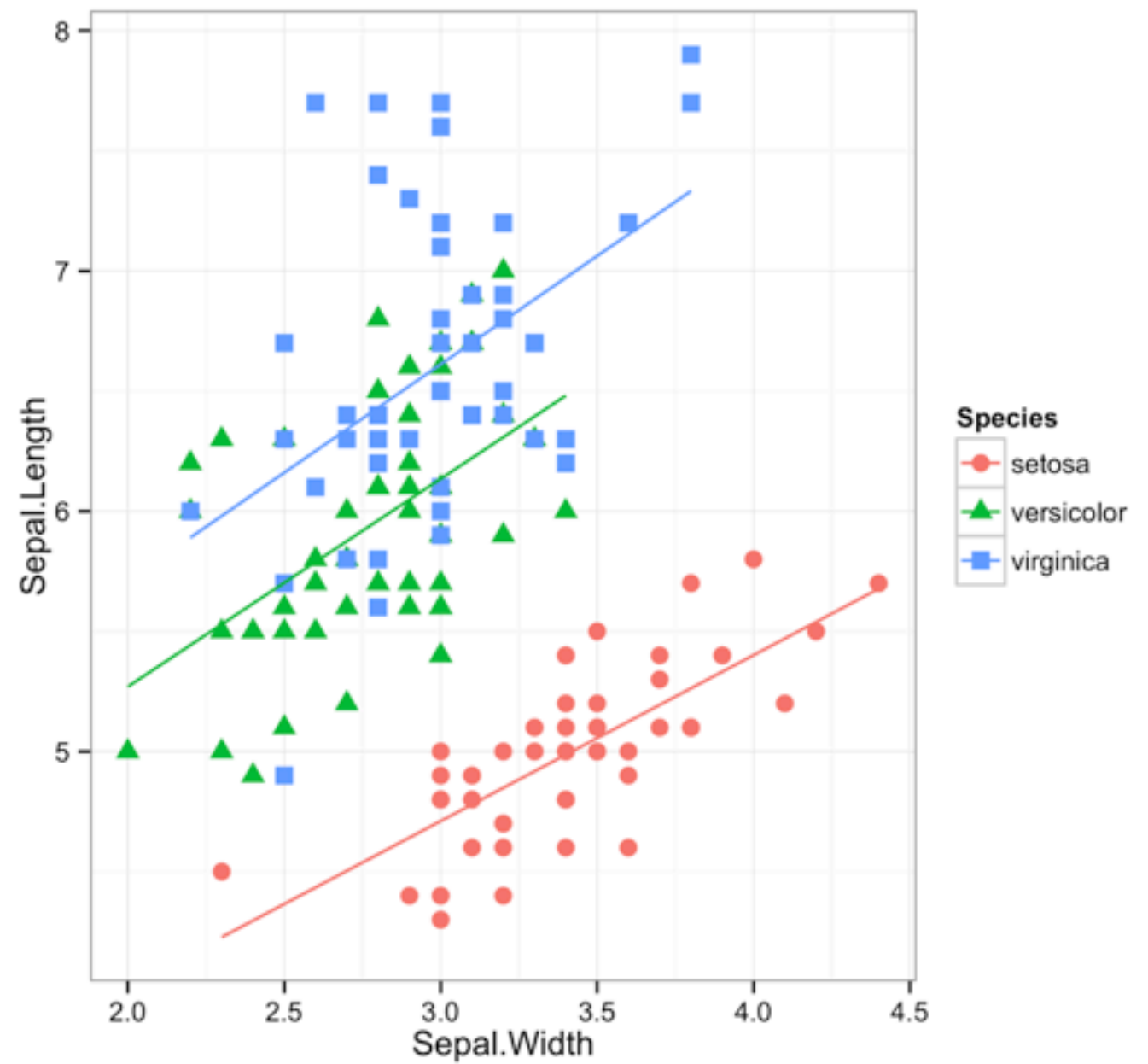
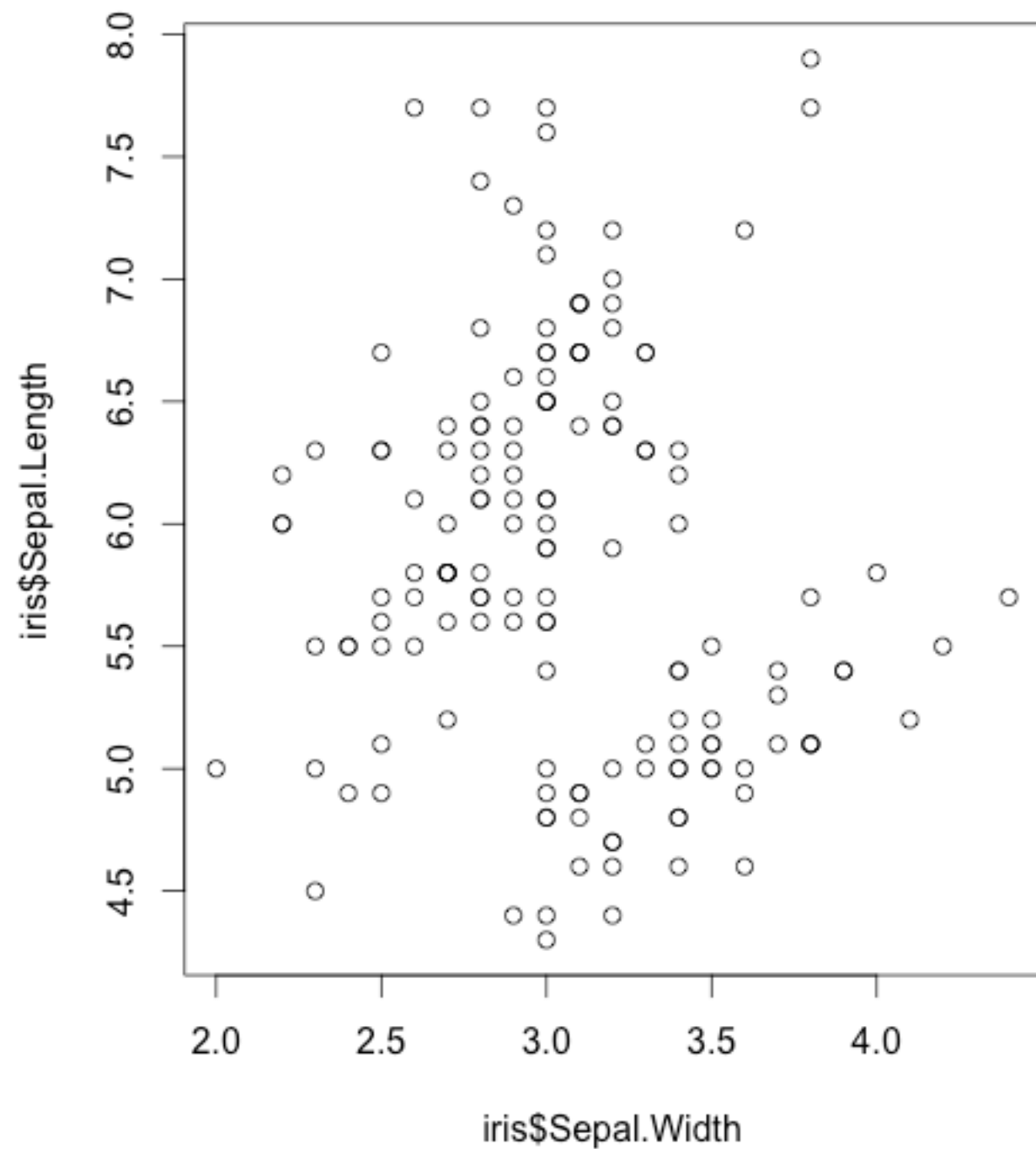
# ggplot2



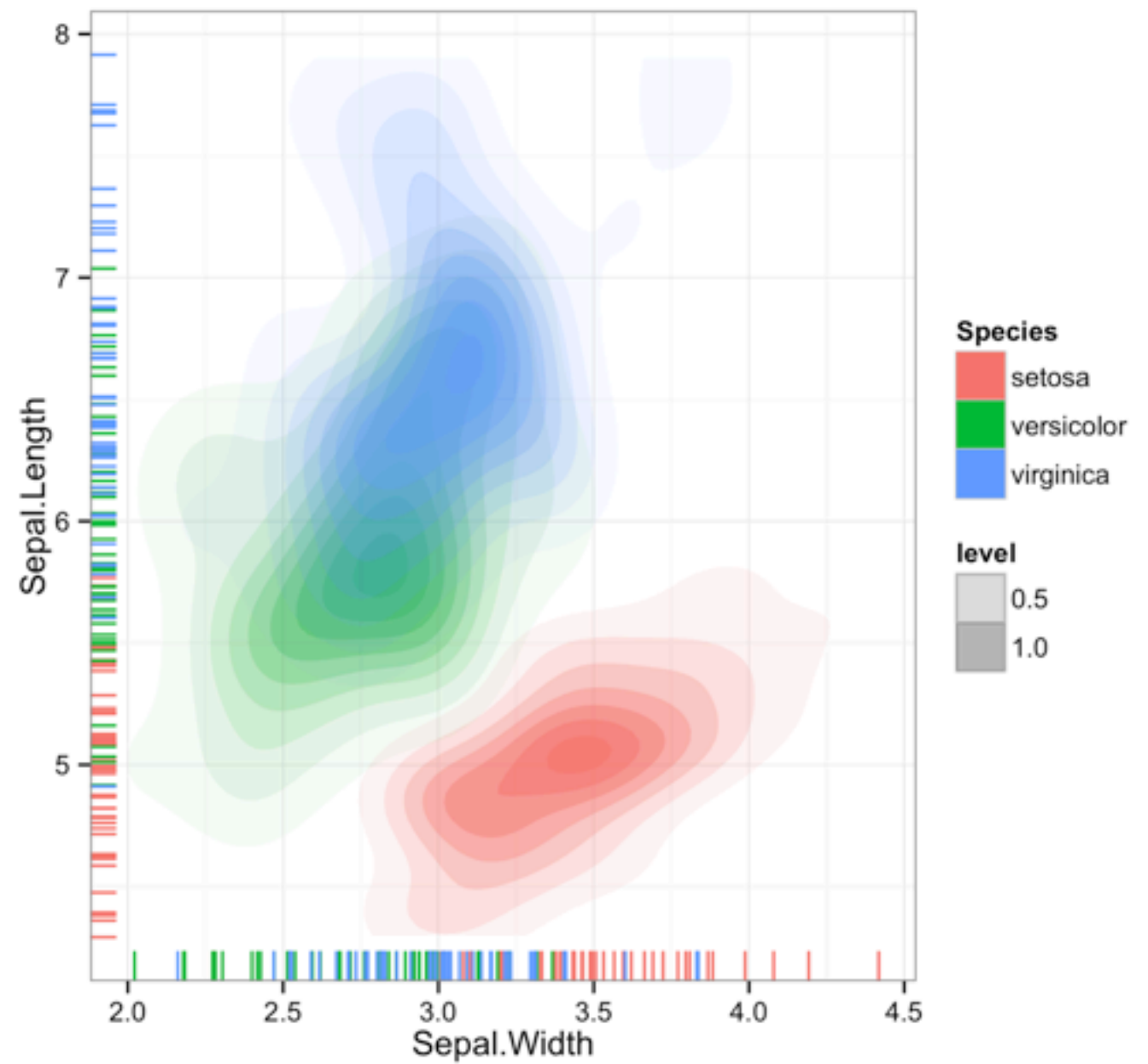
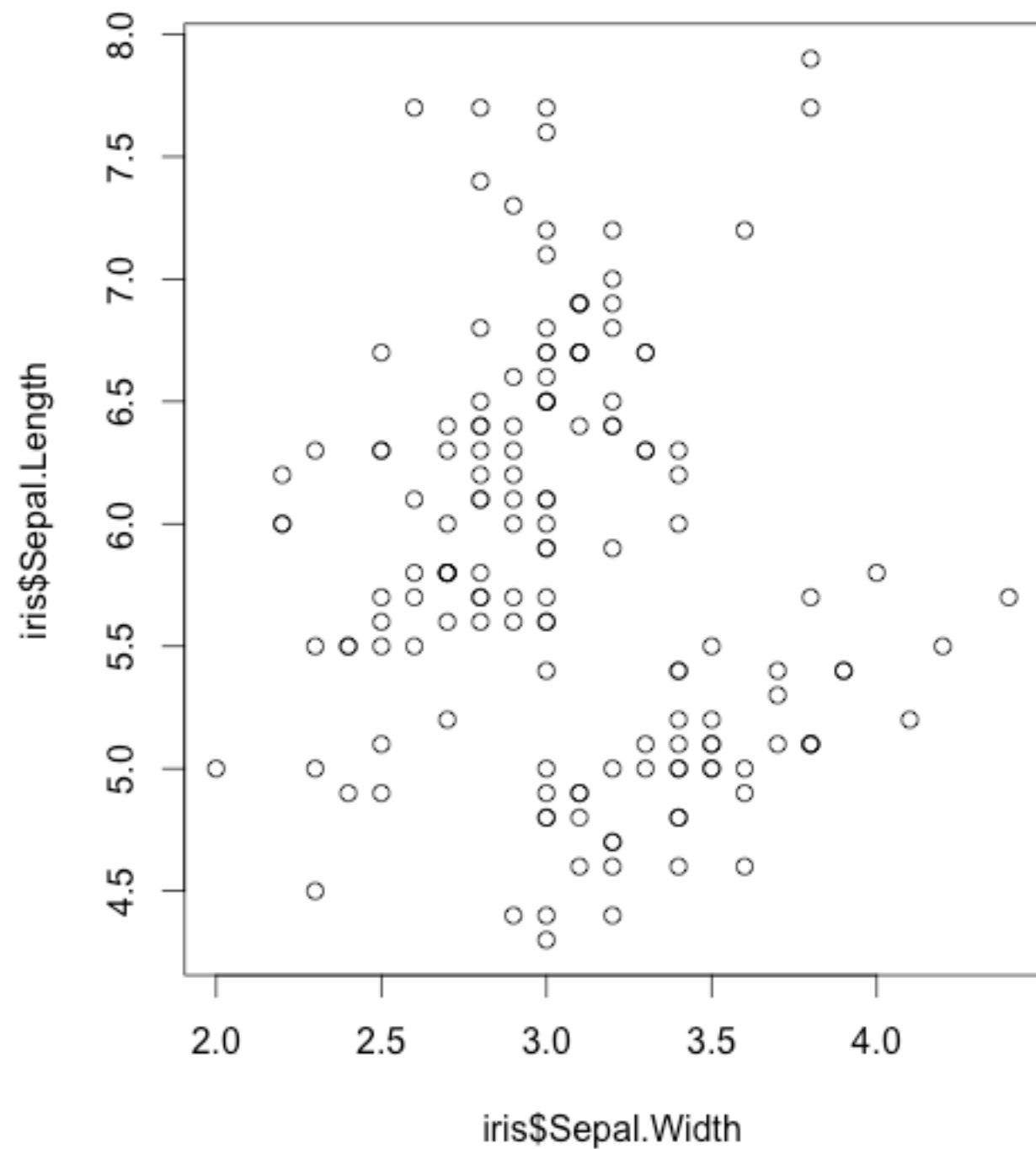
# ggplot2



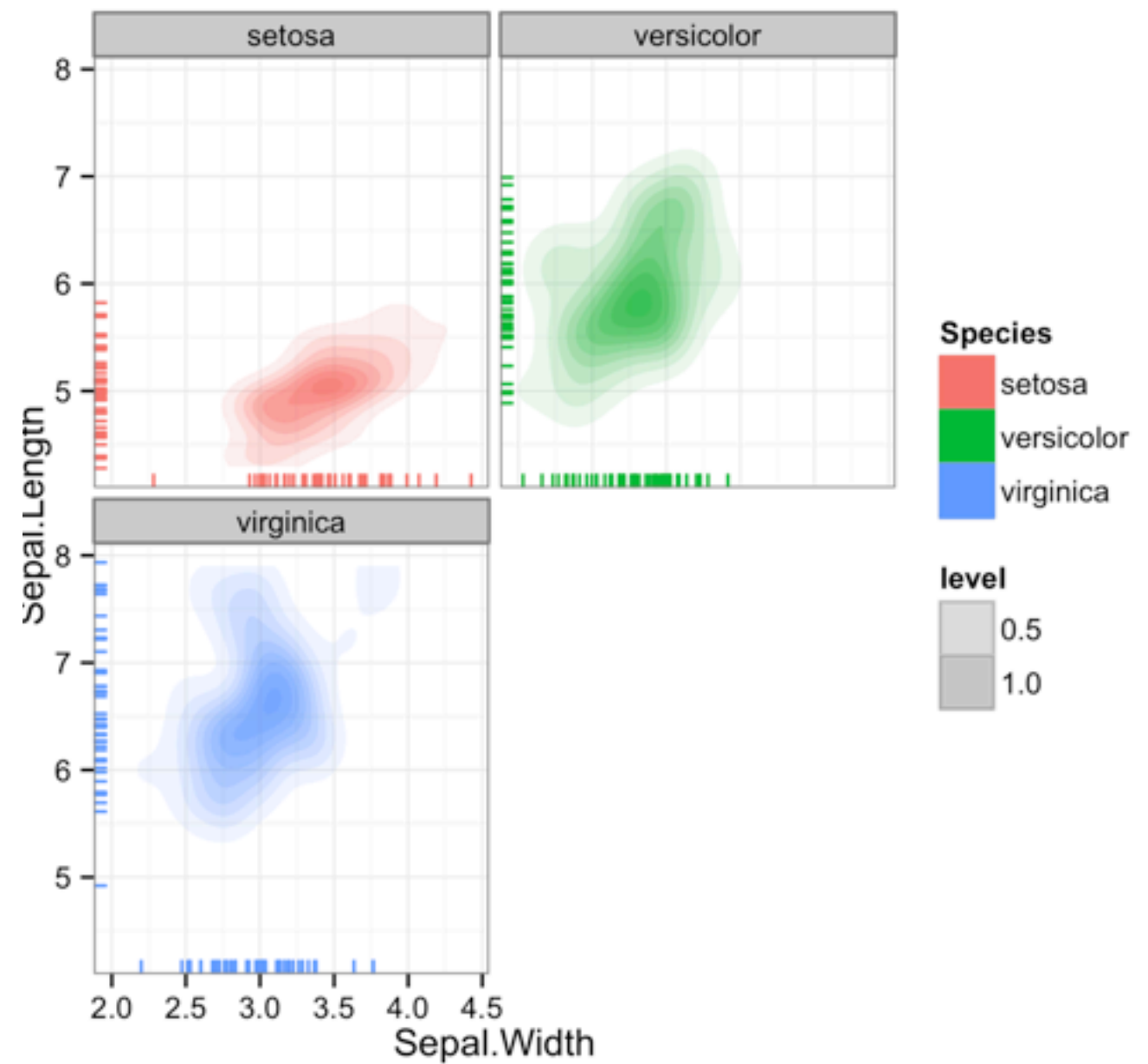
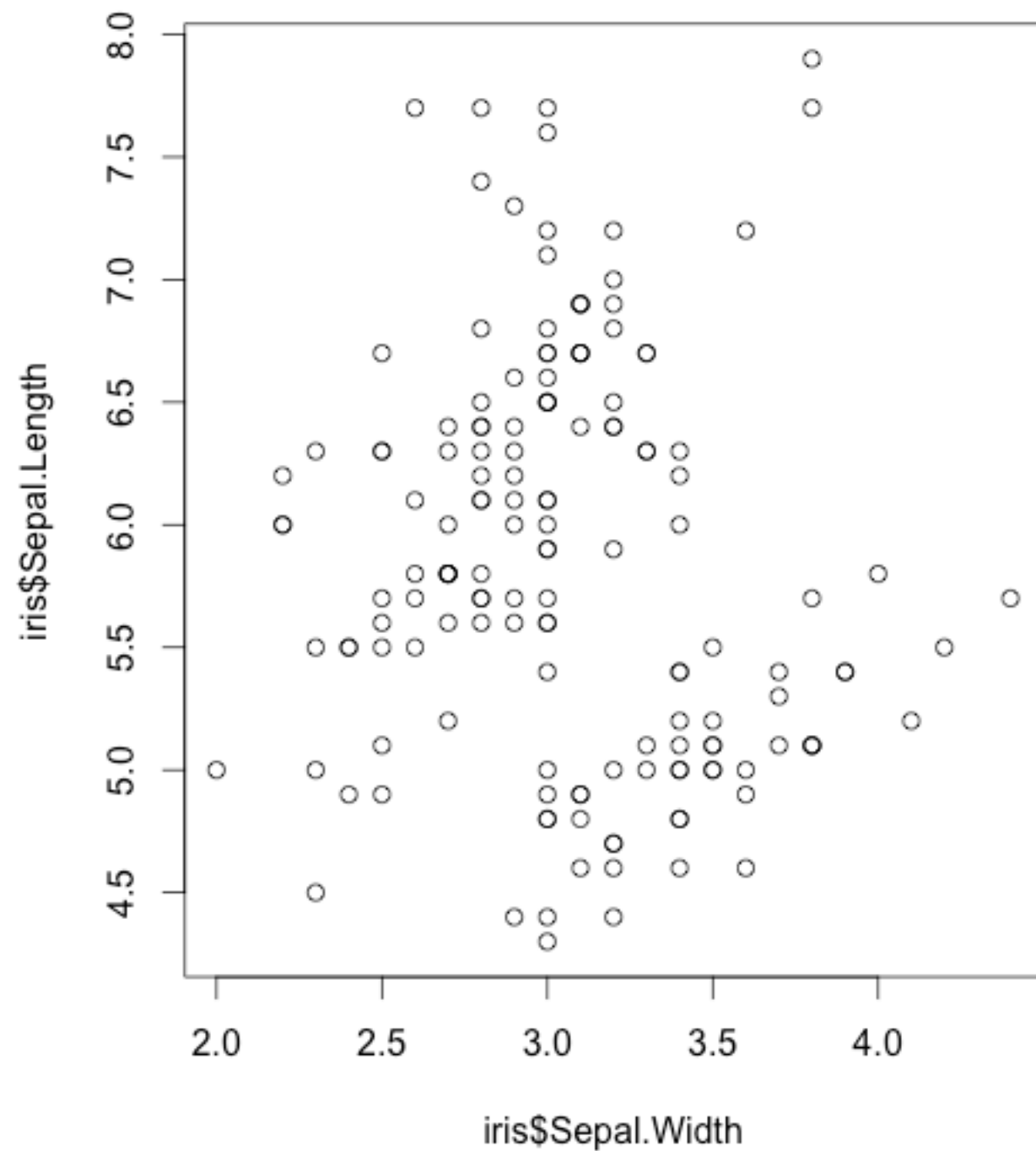
# ggplot2



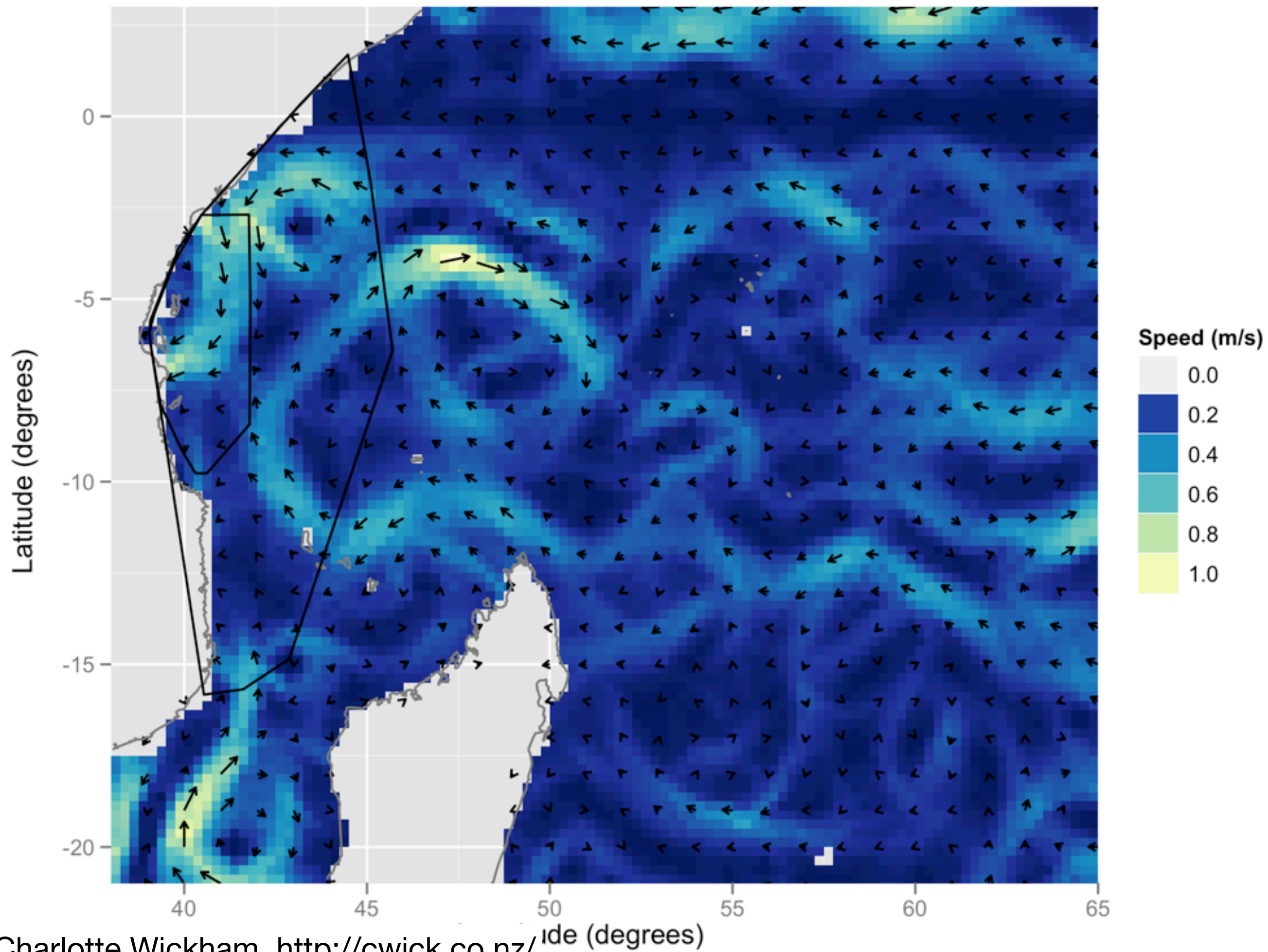
# ggplot2



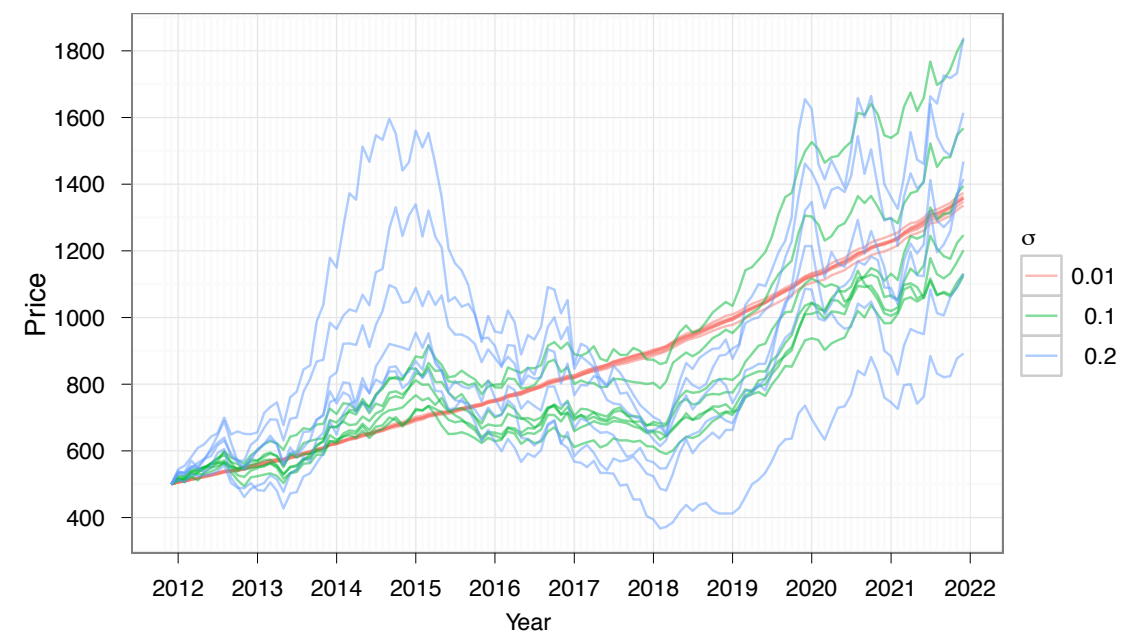
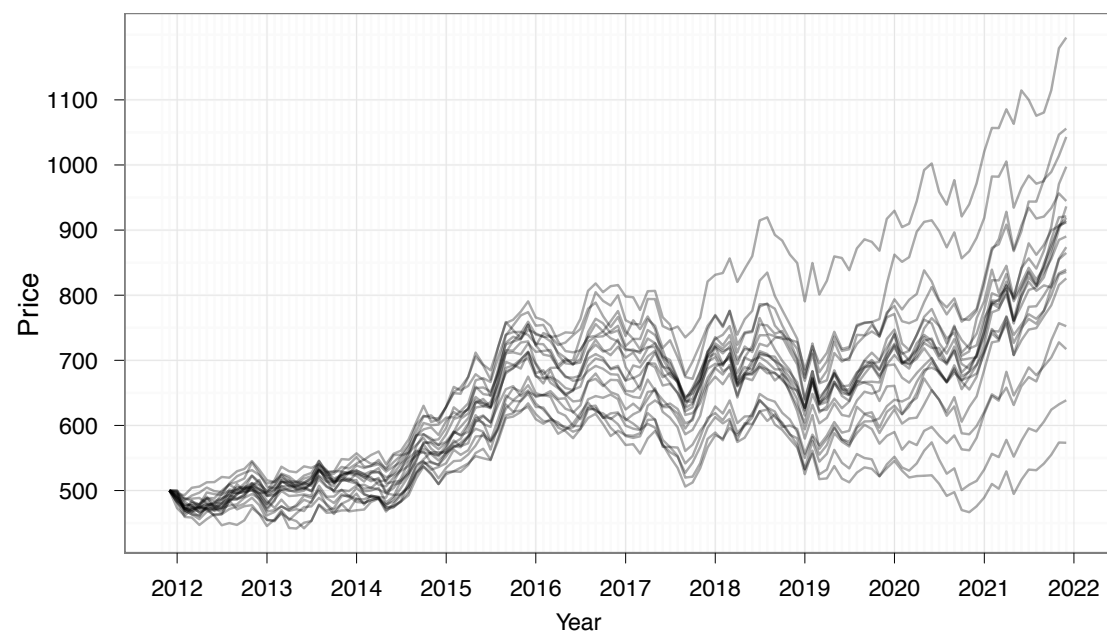
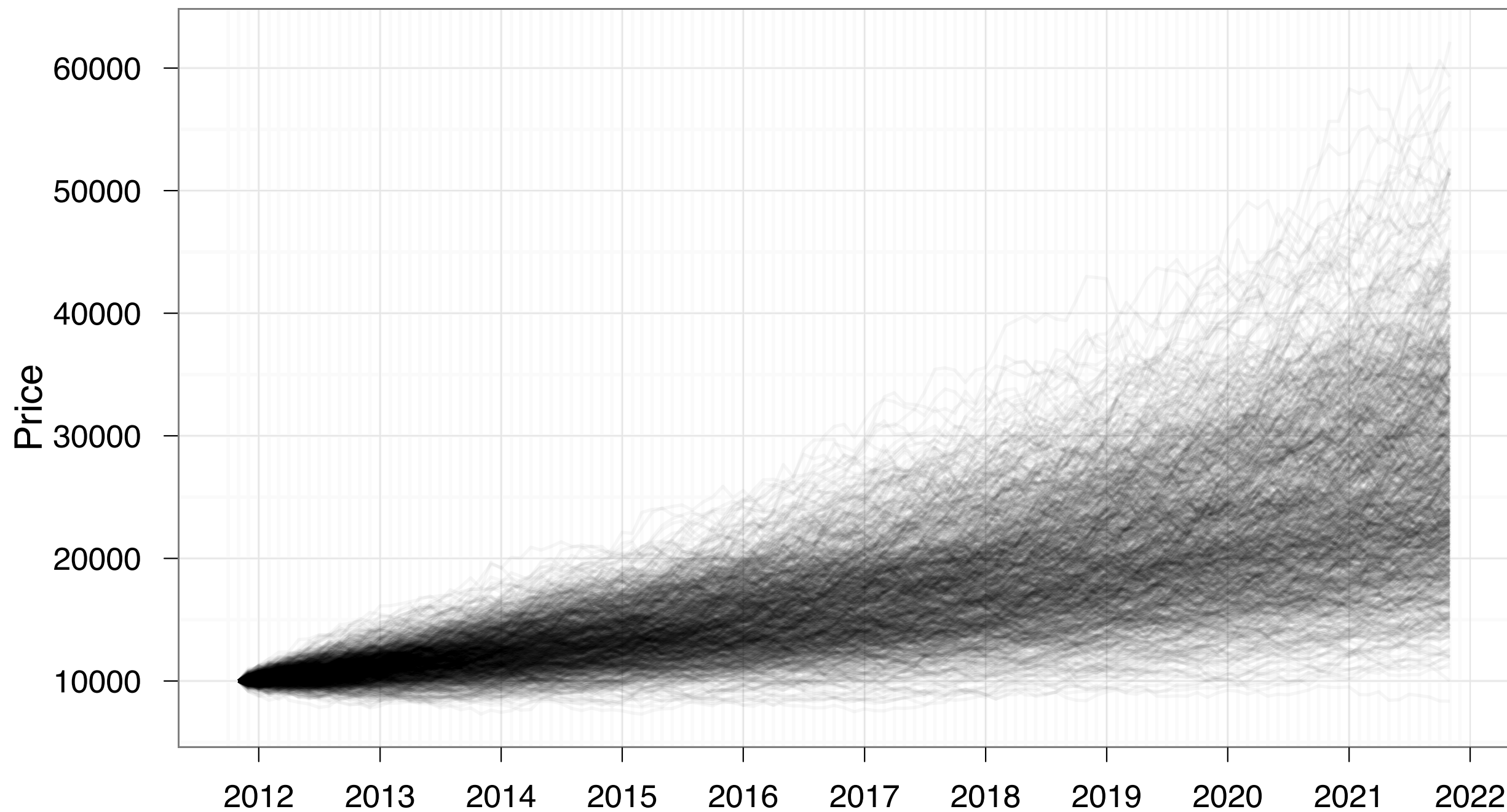
# ggplot2

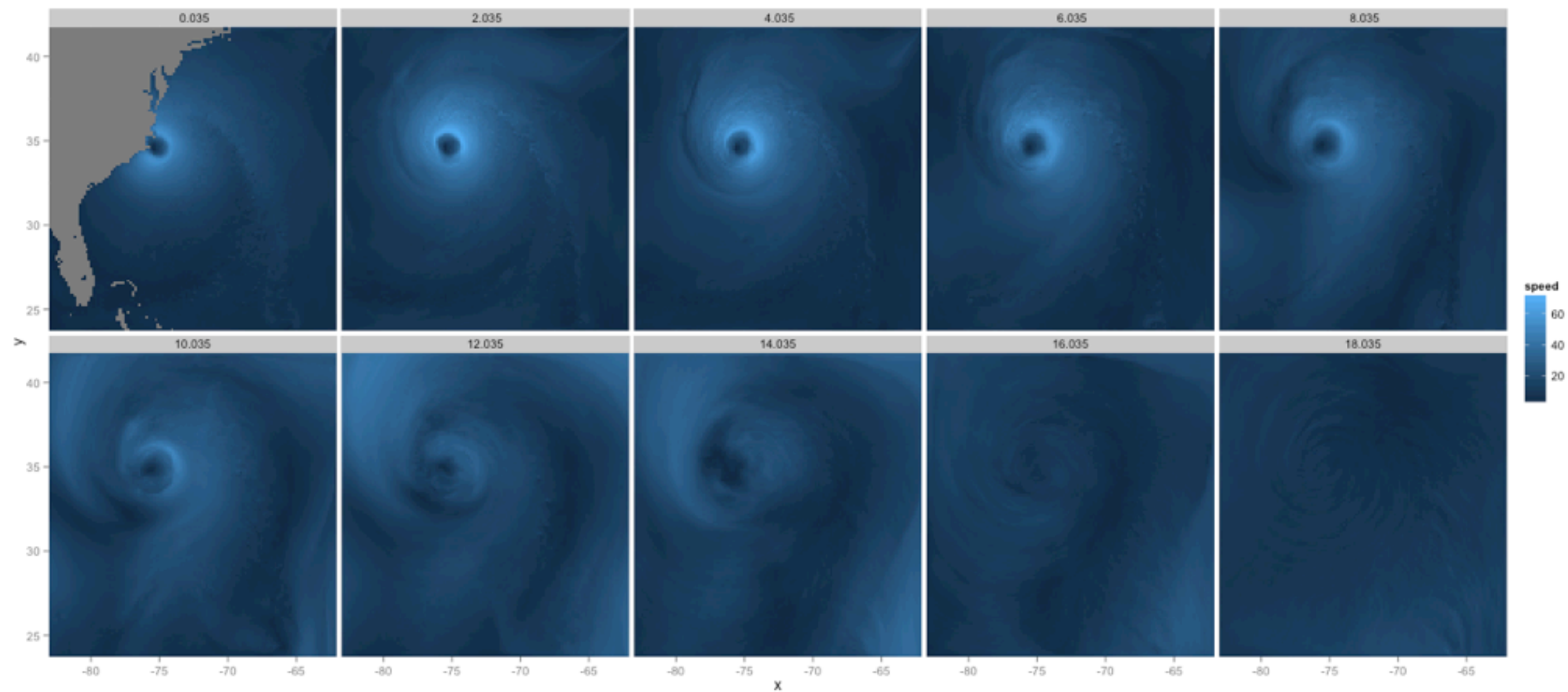


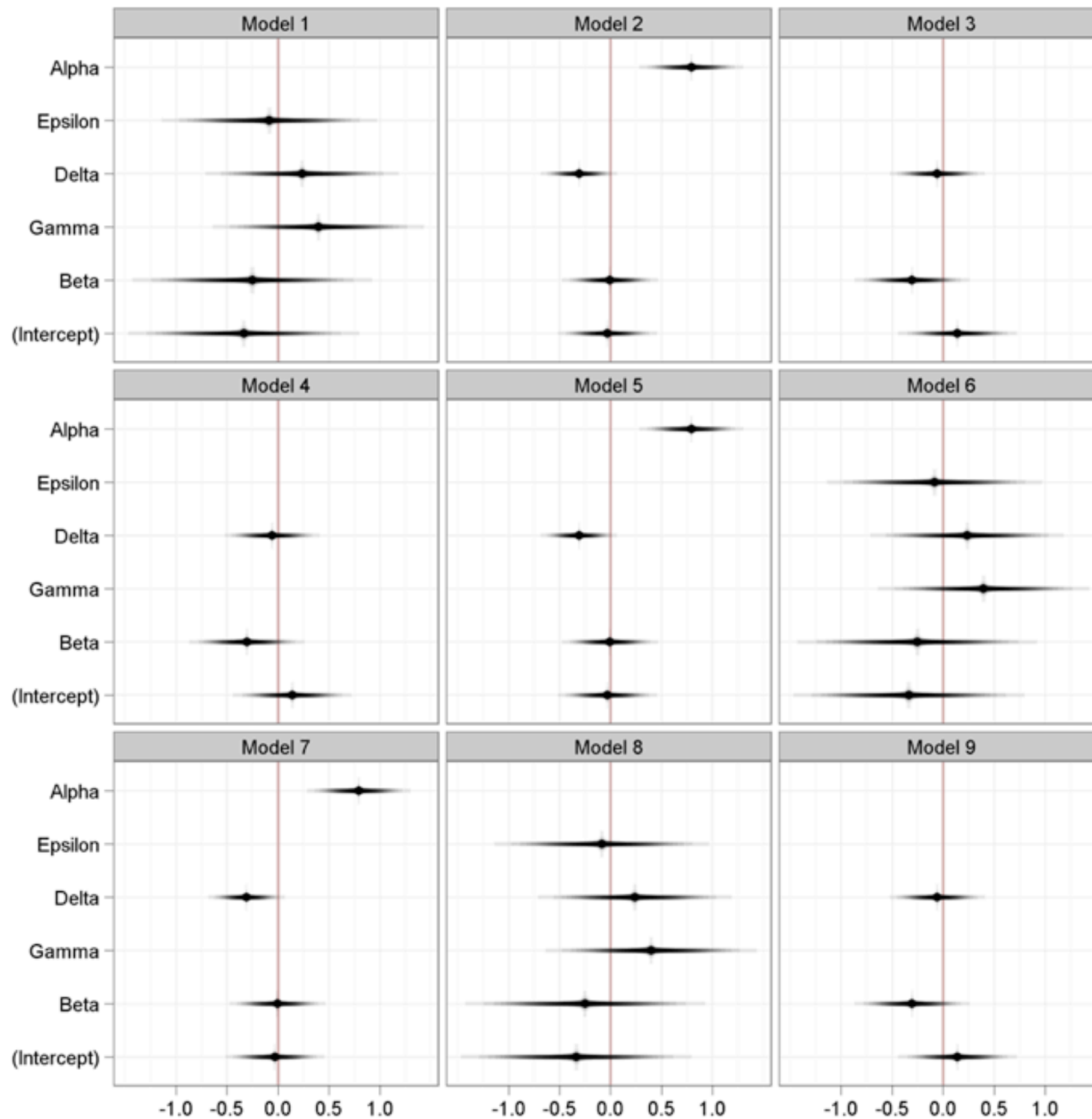




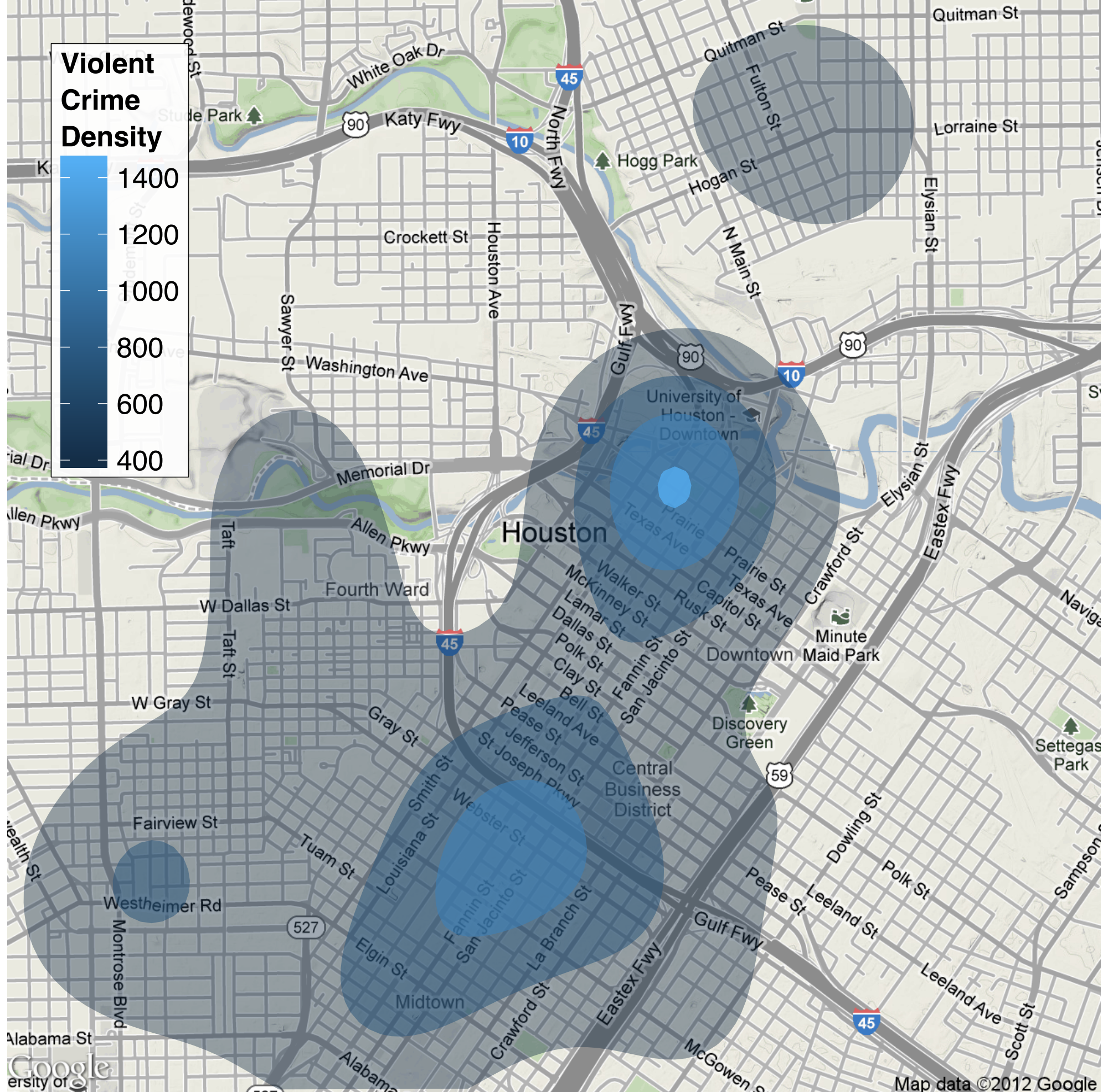














# London Cycle Hire Journeys

Thicker, yellower lines mean more journeys



Data: 3.2 Million Journeys (from TfL)  
Routing: Ollie O'Brien (@oobr) + OpenStreetMap cc-by-sa  
Buildings: OS Opendata Crown Copyright 2011  
Map: James Cheshire (@spatialanalysis)

James Cheshire, <http://bit.ly/xqHhAs>



A picture is not merely worth  
a thousand words, it is much  
more likely to be scrutinized  
than words are to be read.

—John Tukey

# **Diving in: Scatterplots**



# Looking at data with R

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

The mpg data set  
comes in the  
ggplot2 package

```
?mpg  
View(mpg)
```

Always read the  
help page

# Your turn

Make a prediction. What relationship do you expect to see between engine size (displ) and mileage (hwy)?

No peeking ahead!

How can we look at this?

# (quick) plots in R

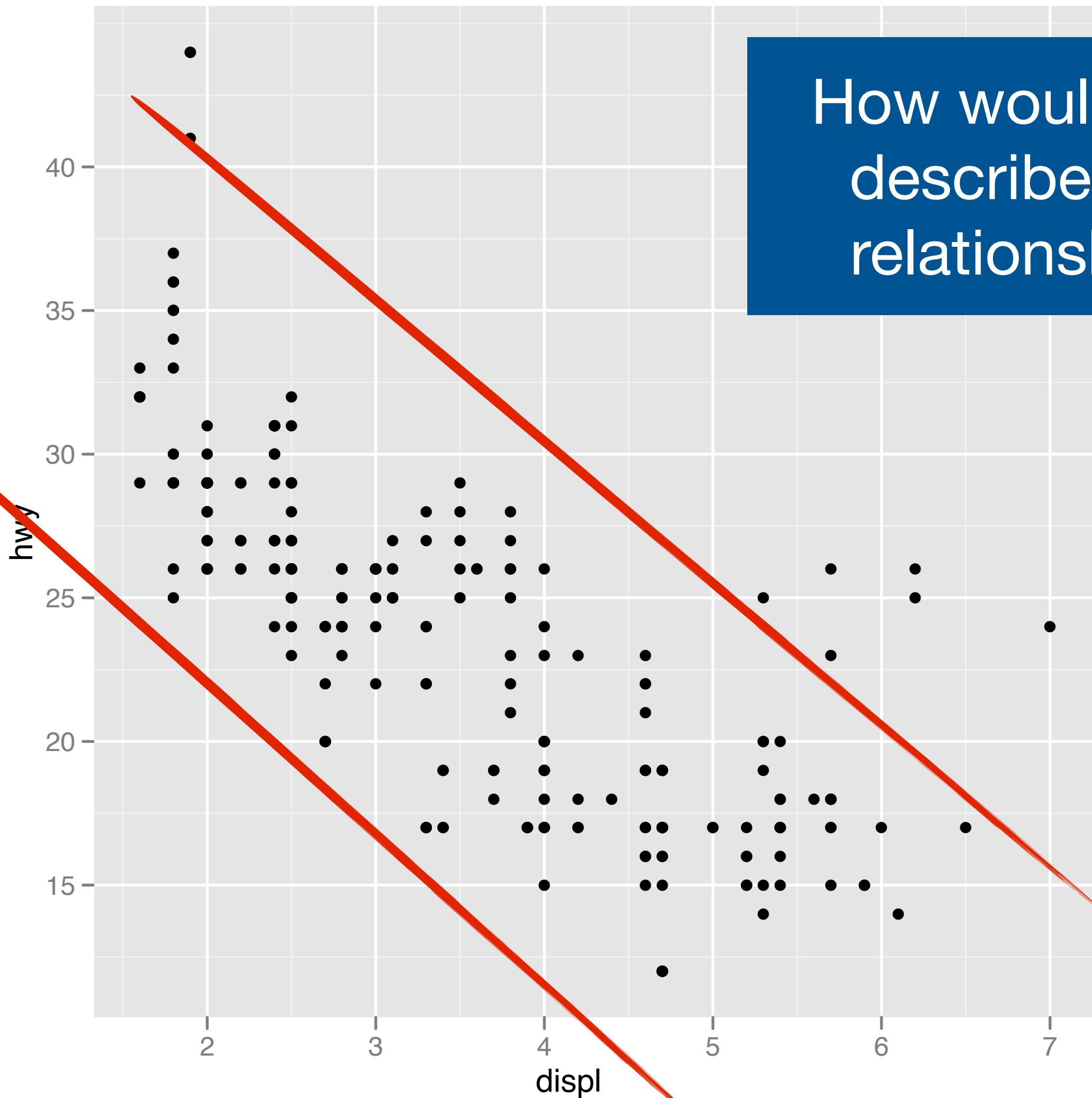
x variable

y variable

data set variables are in

```
qplot(displ, hwy, data = mpg)
```

How would you  
describe this  
relationship?

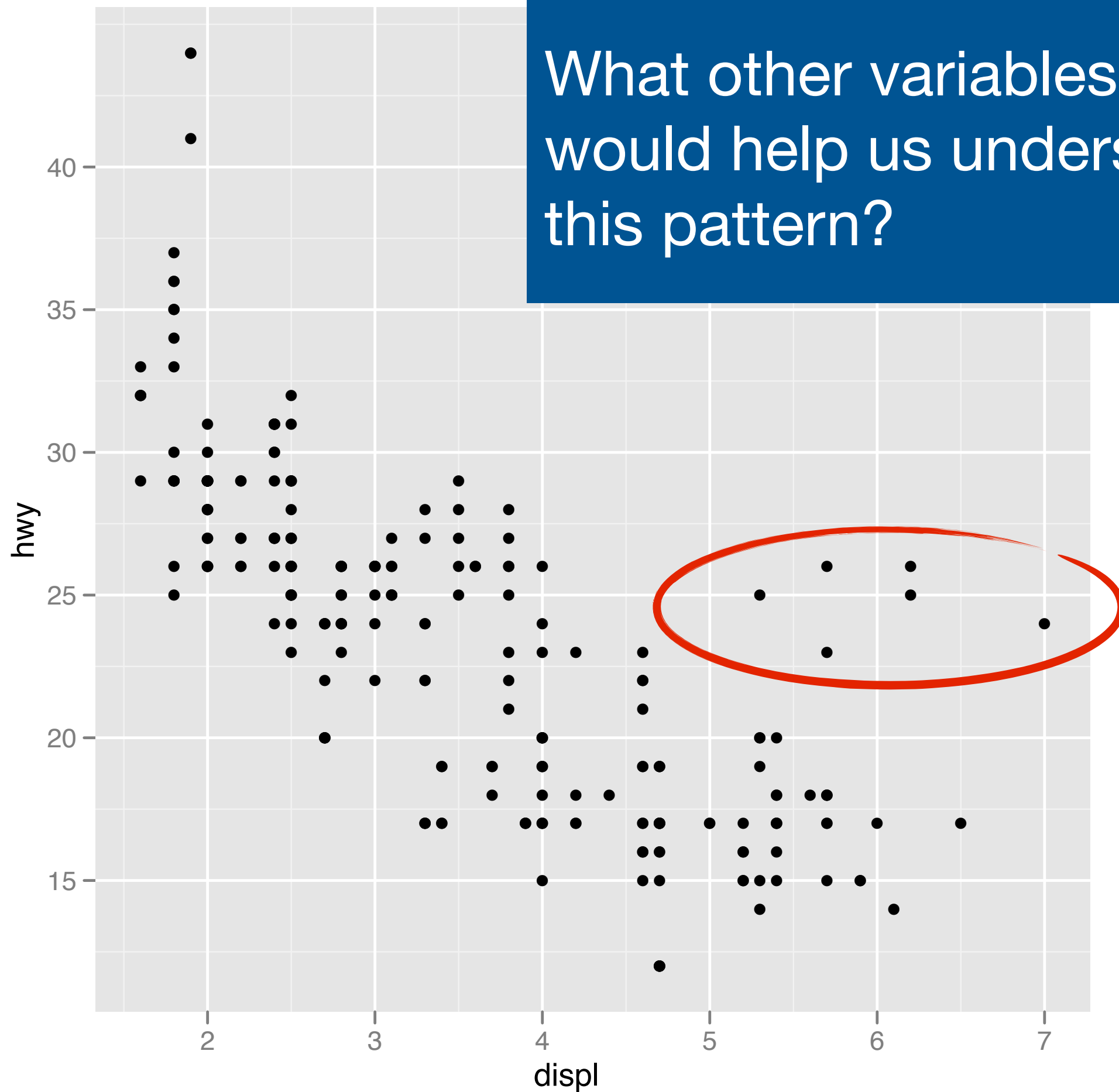


```
qplot(displ, hwy, data = mpg)
```

The greatest value of a picture  
is when it forces us to notice  
what we never expected to  
see.

—John Tukey

What other variables  
would help us understand  
this pattern?



```
qplot(displ, hwy, data = mpg)
```

# Additional variables

Can display additional variables with

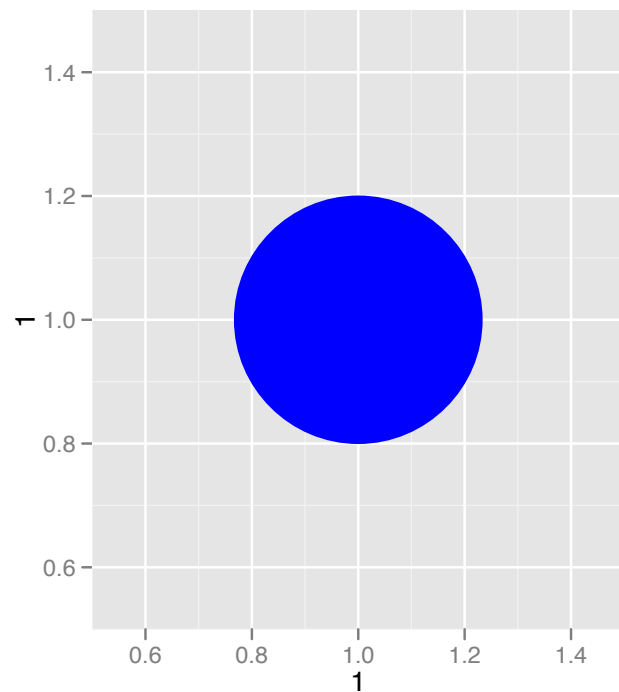
- **aesthetics** (like shape, colour, size), or
- **faceting** (small multiples displaying different subsets)



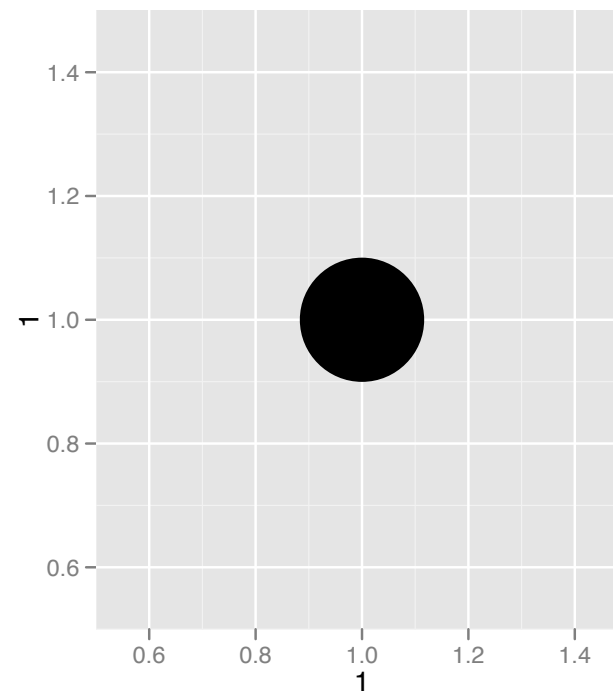
# Aesthetics

# Aesthetics

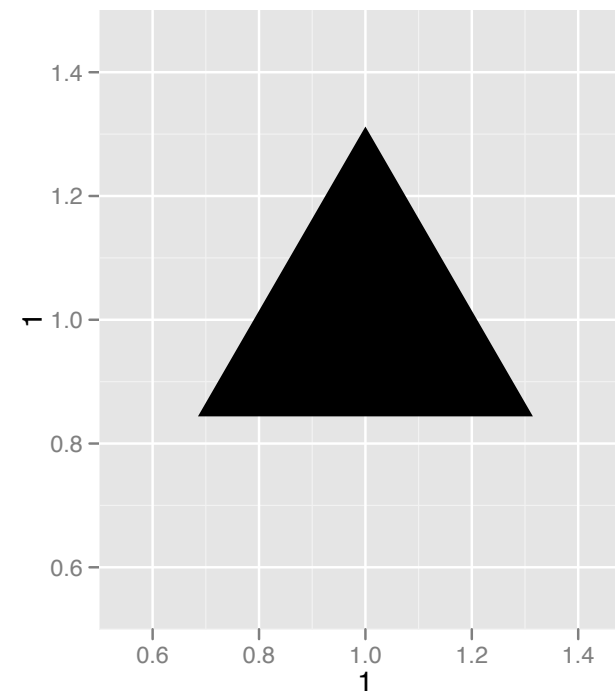
Visual characteristics that can be mapped to data



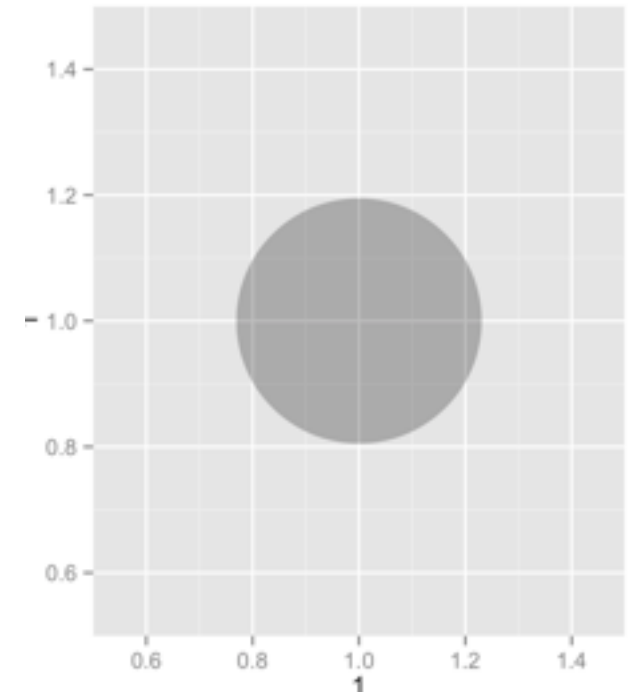
color



size



shape



alpha  
(transparency)

# Aesthetics

aesthetic  
feature

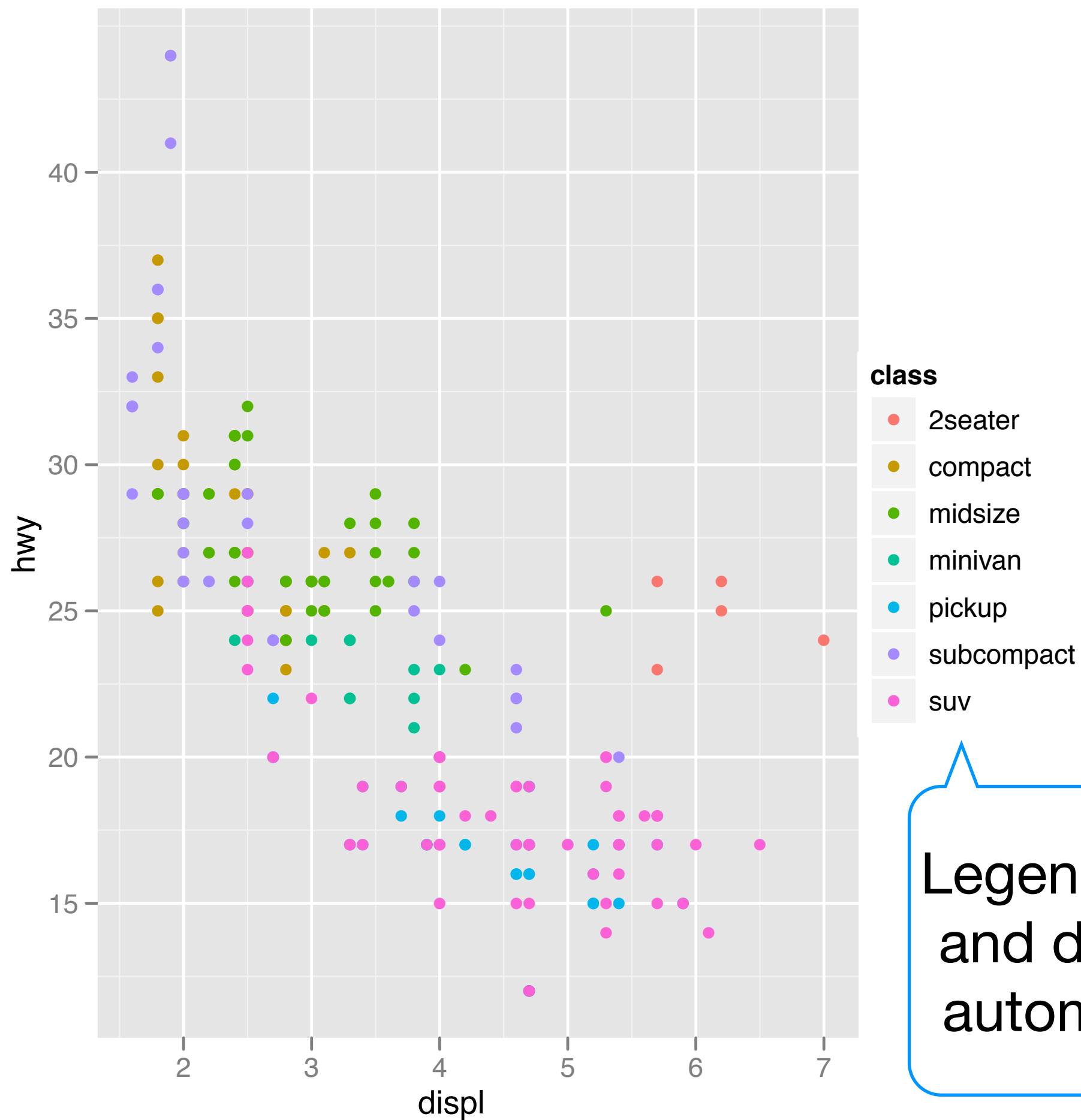
variable to  
map it to

```
qplot(displ, hwy, data = mpg, color = class)
```

```
qplot(displ, hwy, data = mpg, size = class)
```

```
qplot(displ, hwy, data = mpg, shape = class)
```

```
qplot(displ, hwy, data = mpg, alpha = class)
```



```
qplot(displ, hwy, data = mpg, color = class)
```

# Your turn

Add color, size, and shape aesthetics to your graph. Experiment.

Do different things happen for discrete and continuous variables?

What happens when you use more than one aesthetic?

	Discrete	Continuous
Color	Rainbow of colors	Gradient from light blue to dark blue
Size	Discrete size steps	Linear mapping between radius and value
Shape	Different shape for each	Shouldn't (and doesn't) work

# Faceting



# Faceting

Smaller plots that display different subsets of the data.

Also useful for exploring conditional relationships. Useful for large data.

# Your turn

```
qplot(displ, hwy, data = mpg) +  
facet_grid(. ~ cyl)
```

```
qplot(displ, hwy, data = mpg) +  
facet_grid(drv ~ .)
```

```
qplot(displ, hwy, data = mpg) +  
facet_grid(drv ~ cyl)
```

```
qplot(displ, hwy, data = mpg) +  
facet_wrap(~ class)
```

# Summary

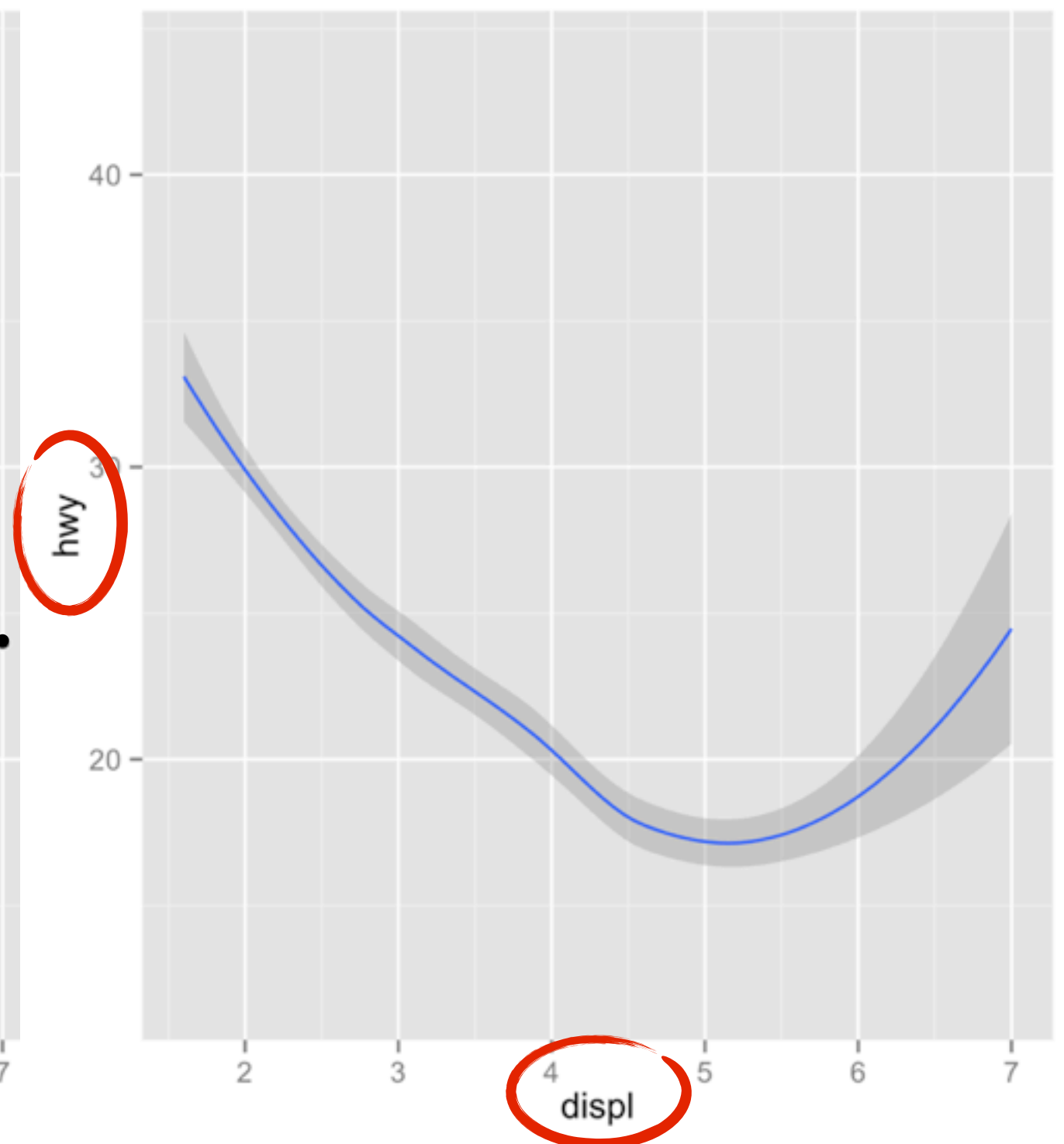
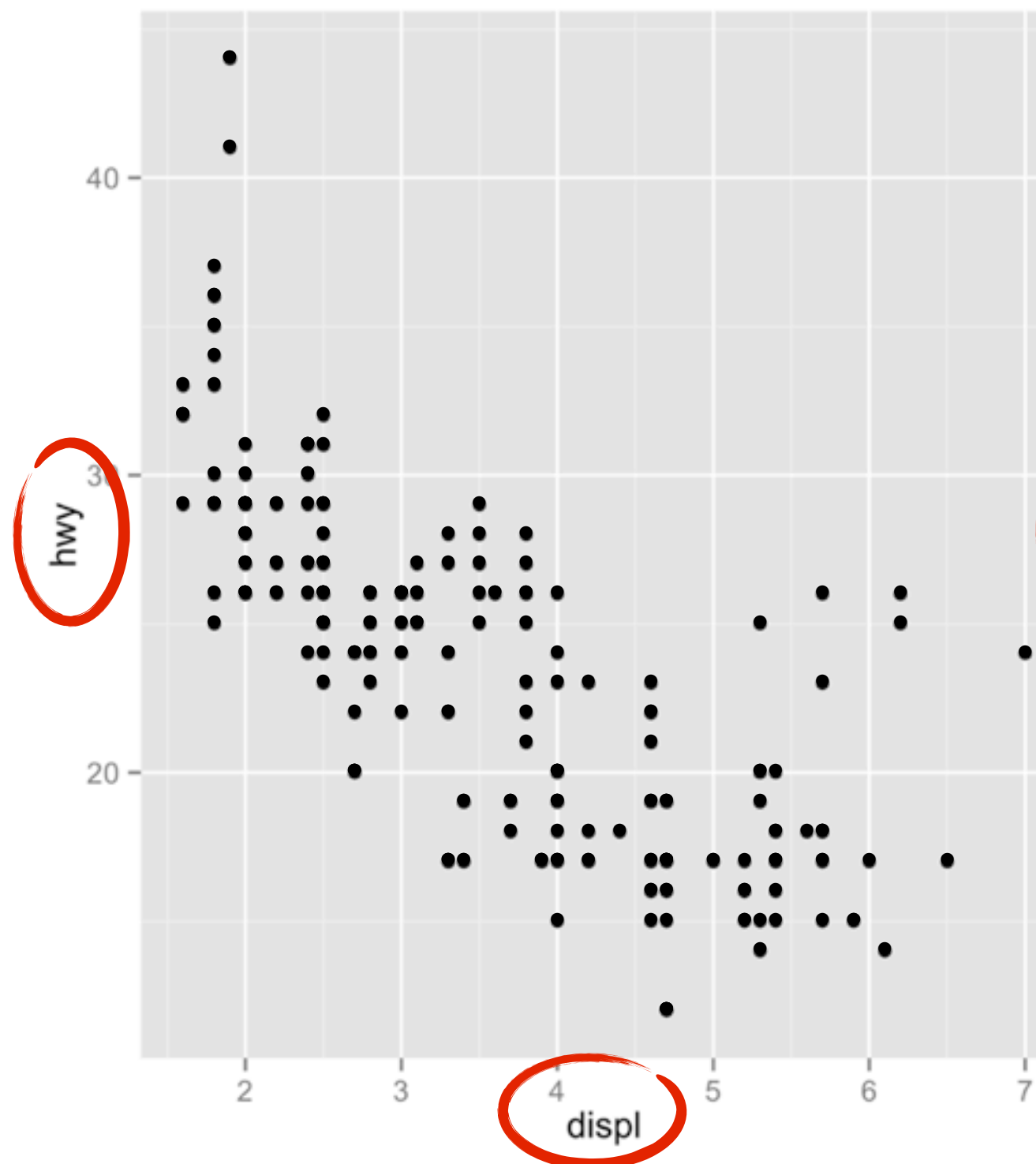
`facet_grid()`: 2d grid, rows ~ cols, . for no split

`facet_wrap()`: 1d ribbon wrapped into 2d

# Geoms

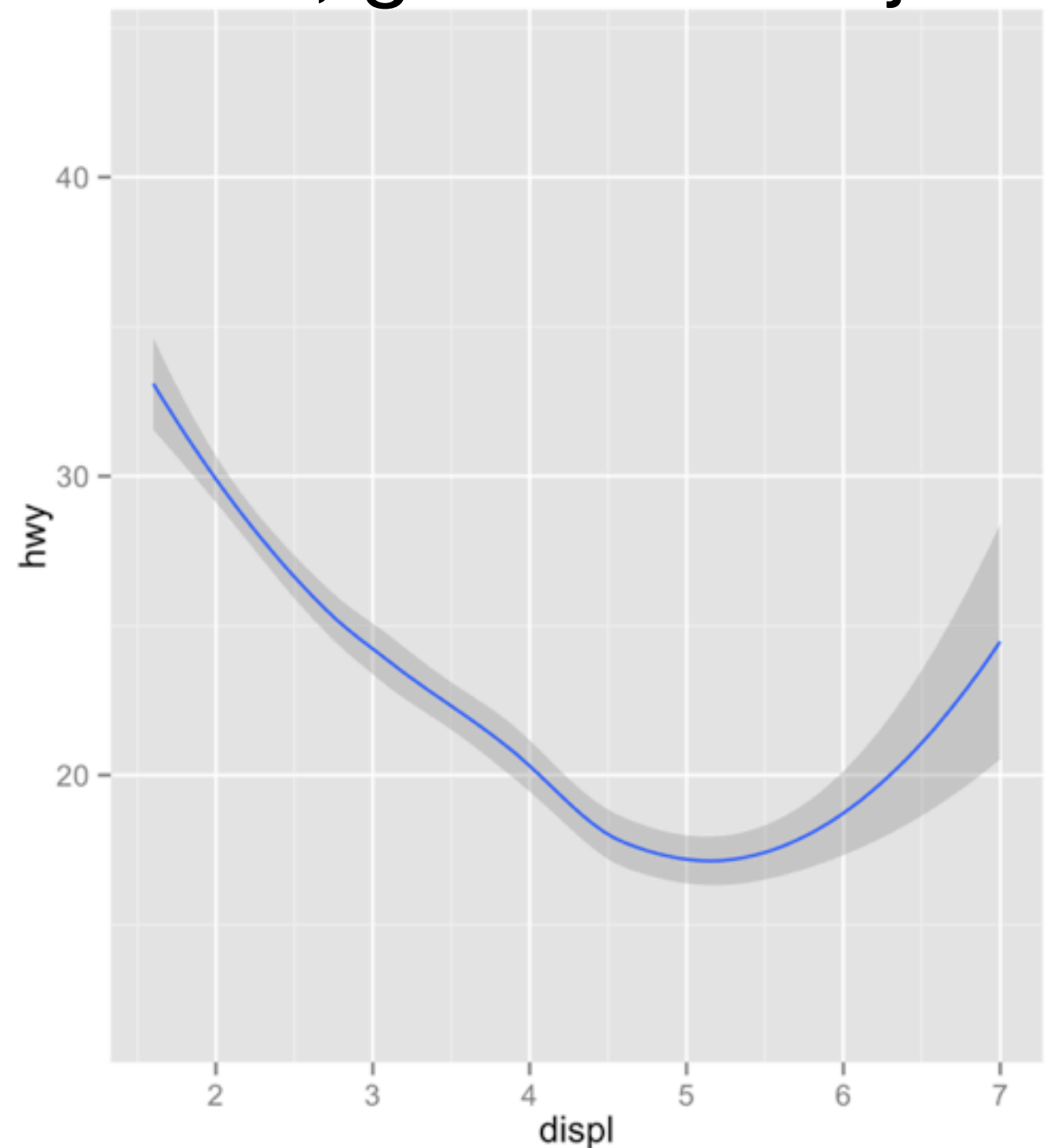
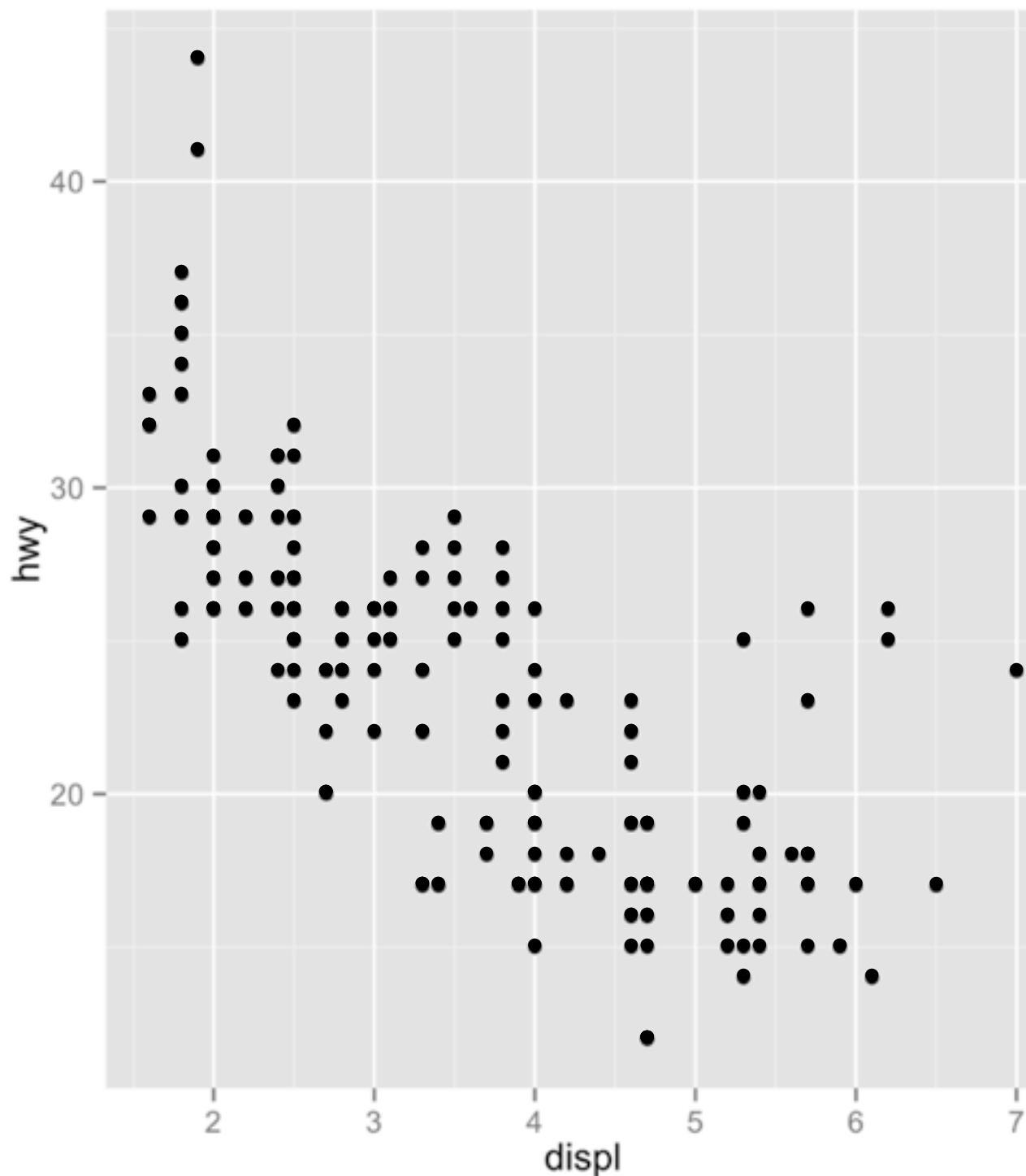
How are these plots similar?

Same: x var, y var, data



How are these plots different?

Different: "type" of plot  
i.e, what plot draws  
i.e, geometric object



# Geometric object

the "type" of graph, or  
what the graph draws

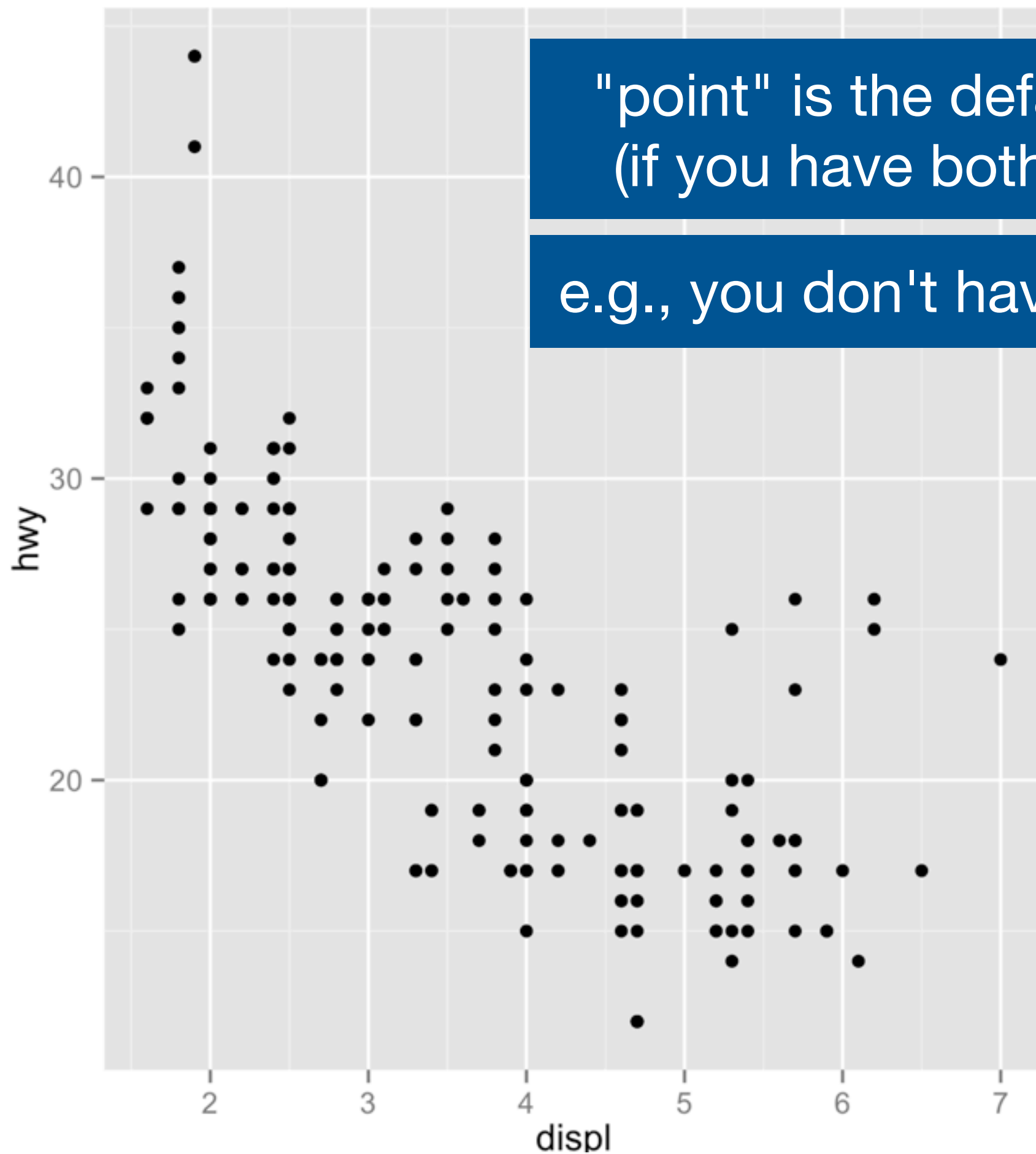
x variable

y variable

data set  
variables are in

type of plot

```
qplot(displ, hwy, data = mpg) geom = "smooth")
```

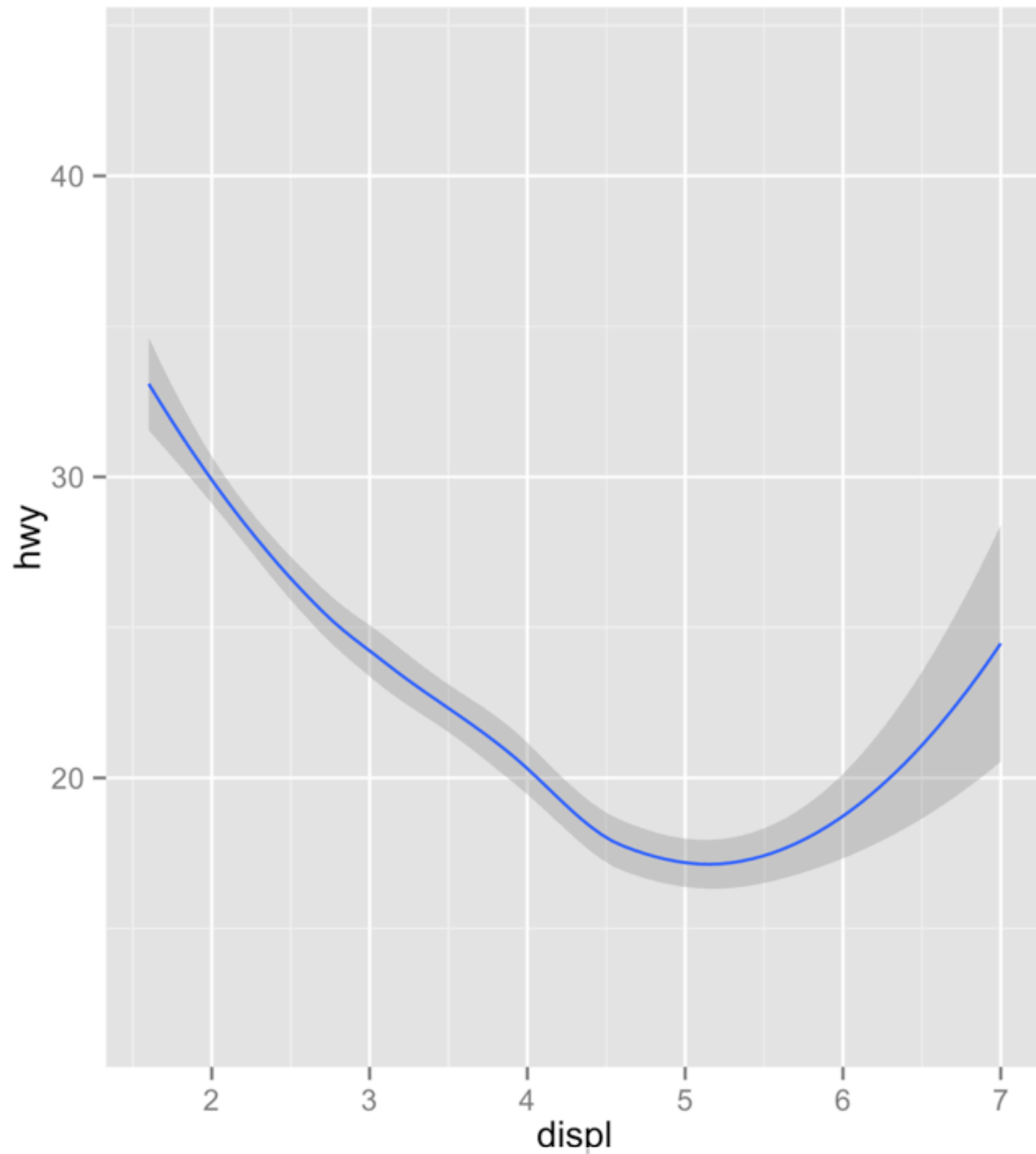


"point" is the default geom  
(if you have both x and y)

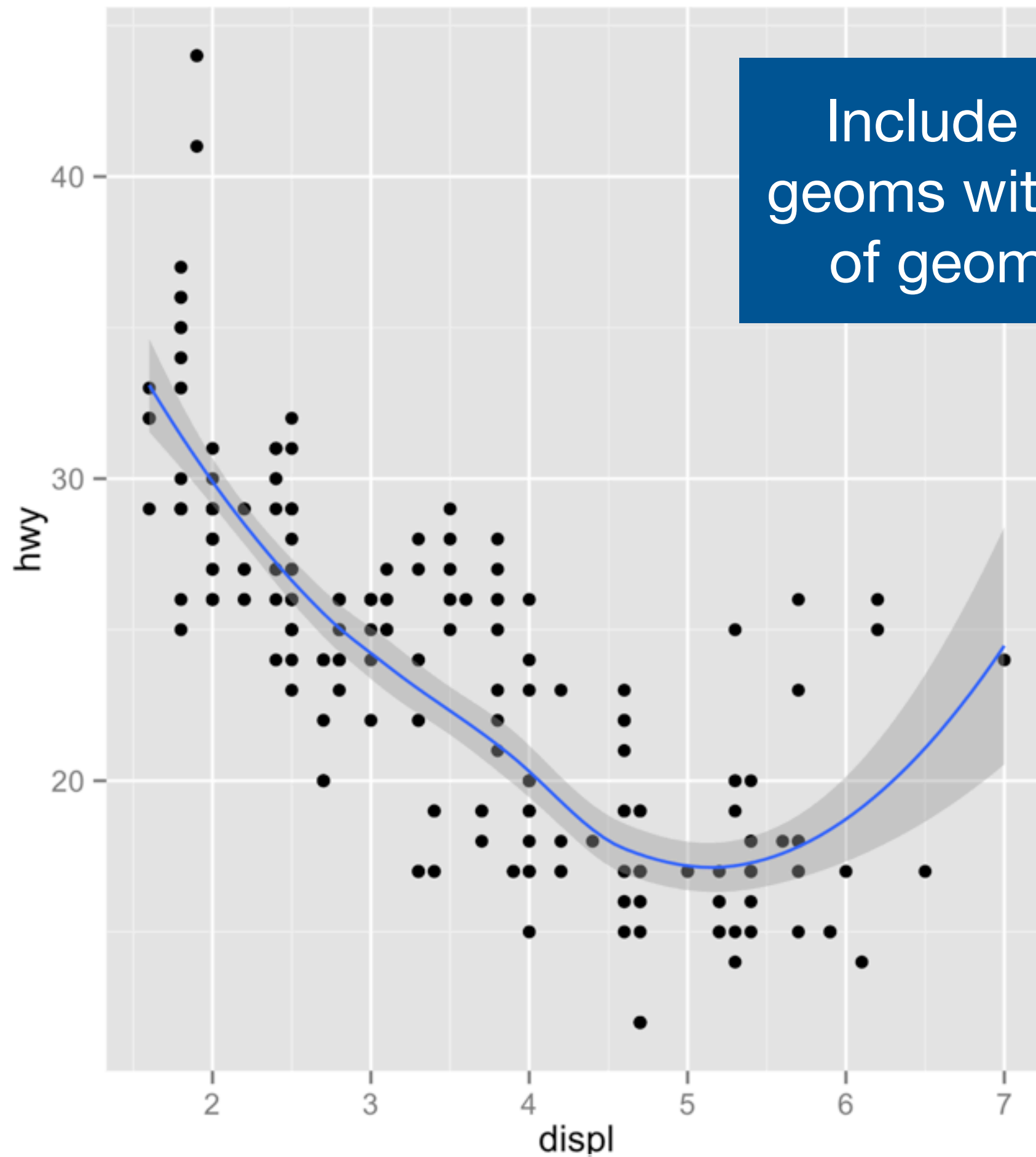
e.g., you don't have to type it

```
qplot(displ, hwy, data = mpg) geom = "point")
```





```
qplot(displ, hwy, data = mpg, geom = "smooth")
```



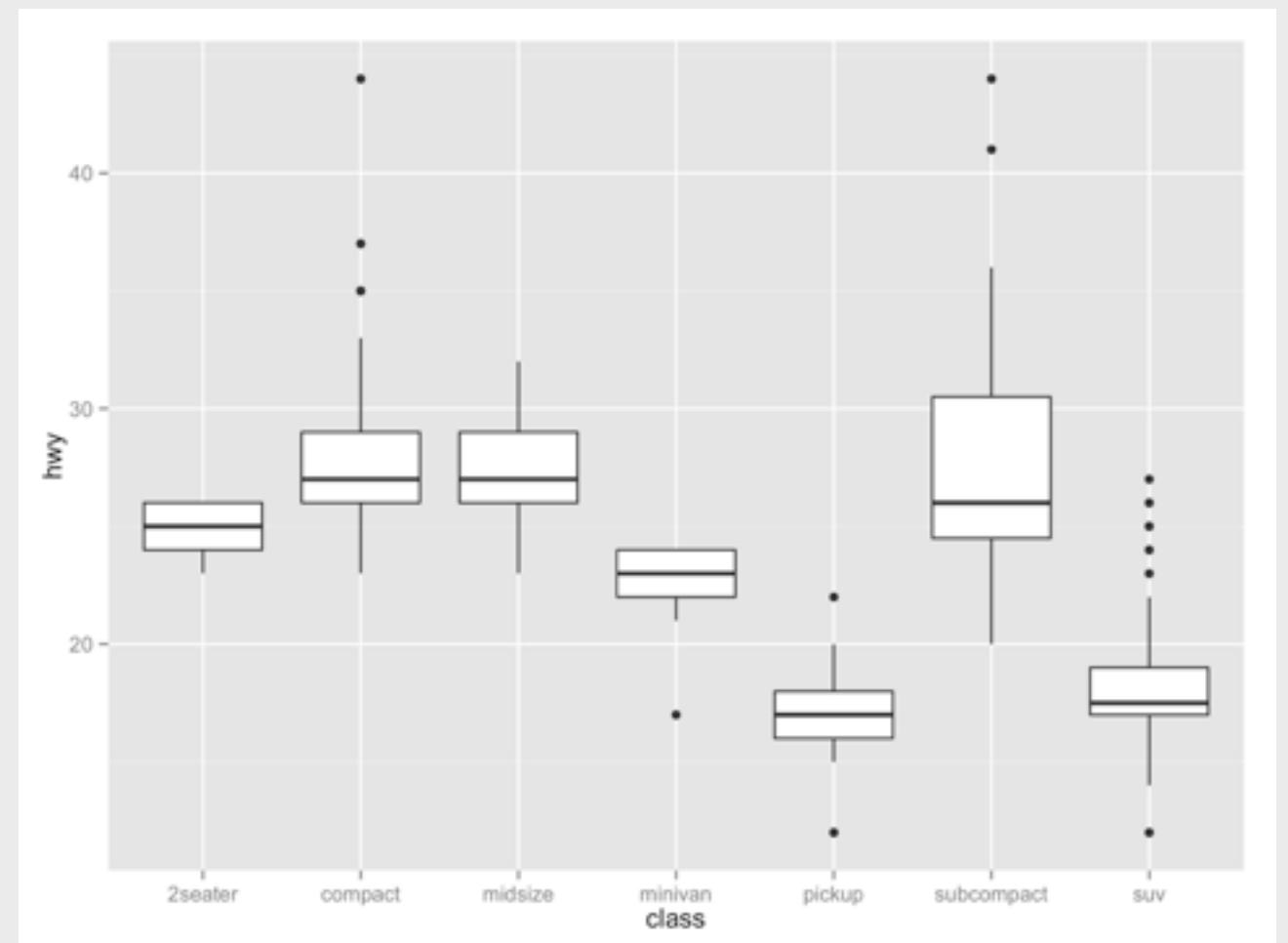
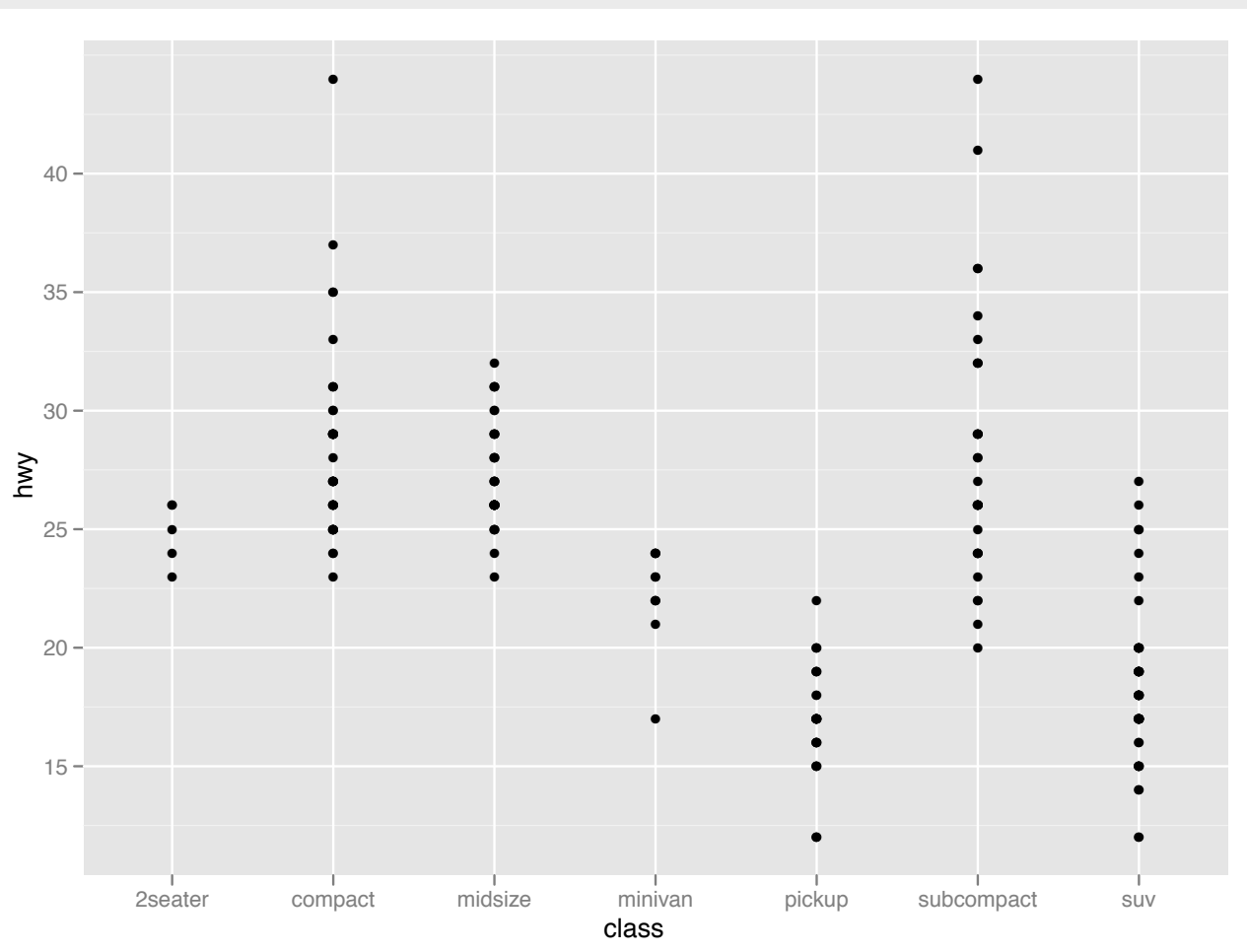
Include multiple  
geoms with a vector  
of geom names

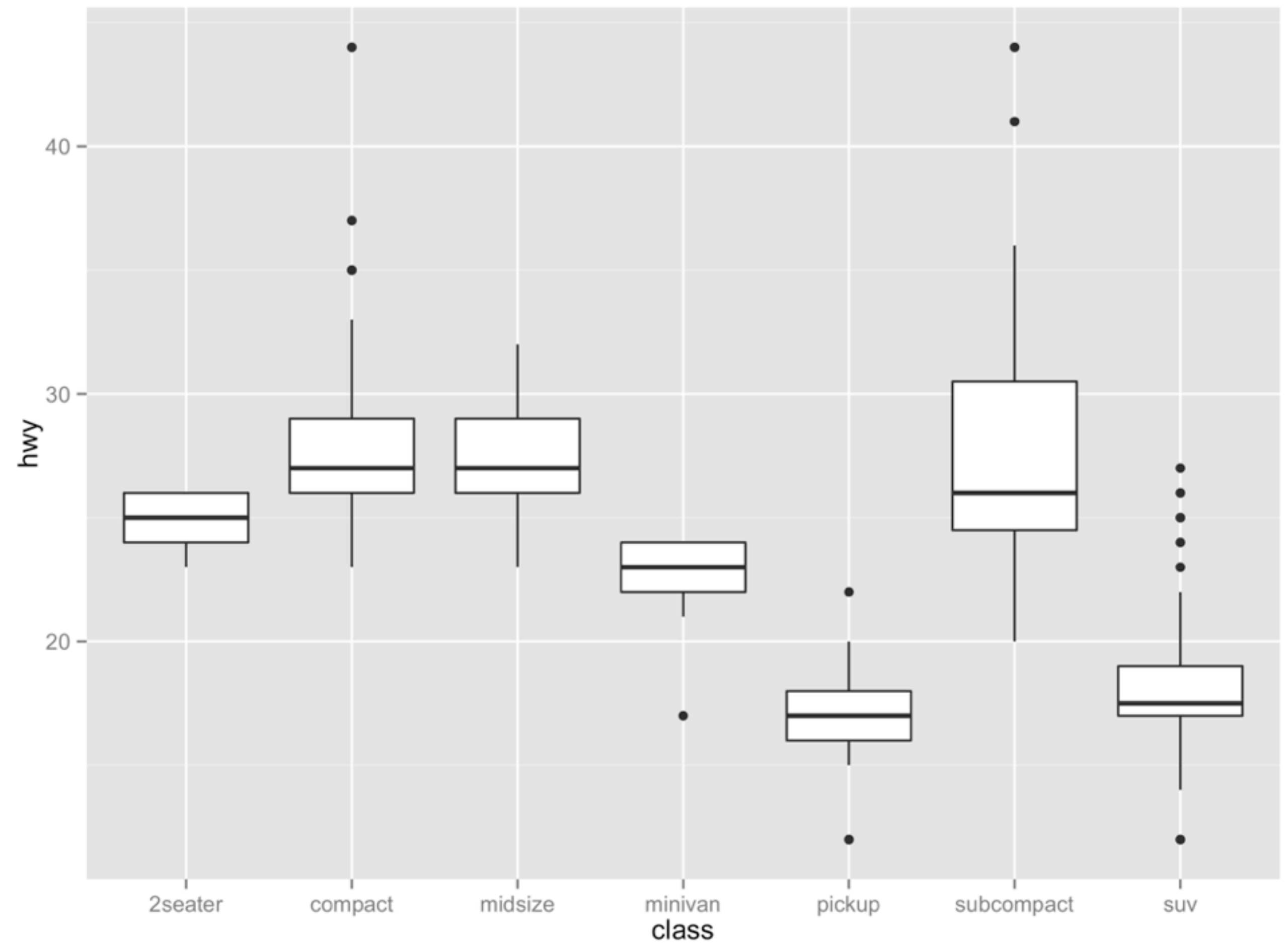
```
qplot(displ, hwy, data = mpg, geom = c("point", "smooth"))
```

# Your turn

How would you replace this scatterplot with one that draws boxplots? Try out your best guess.

```
qplot(class, hwy, data = mpg)
```





```
qplot(class, hwy, data = mpg, geom = "boxplot")
```

# **Diamonds**

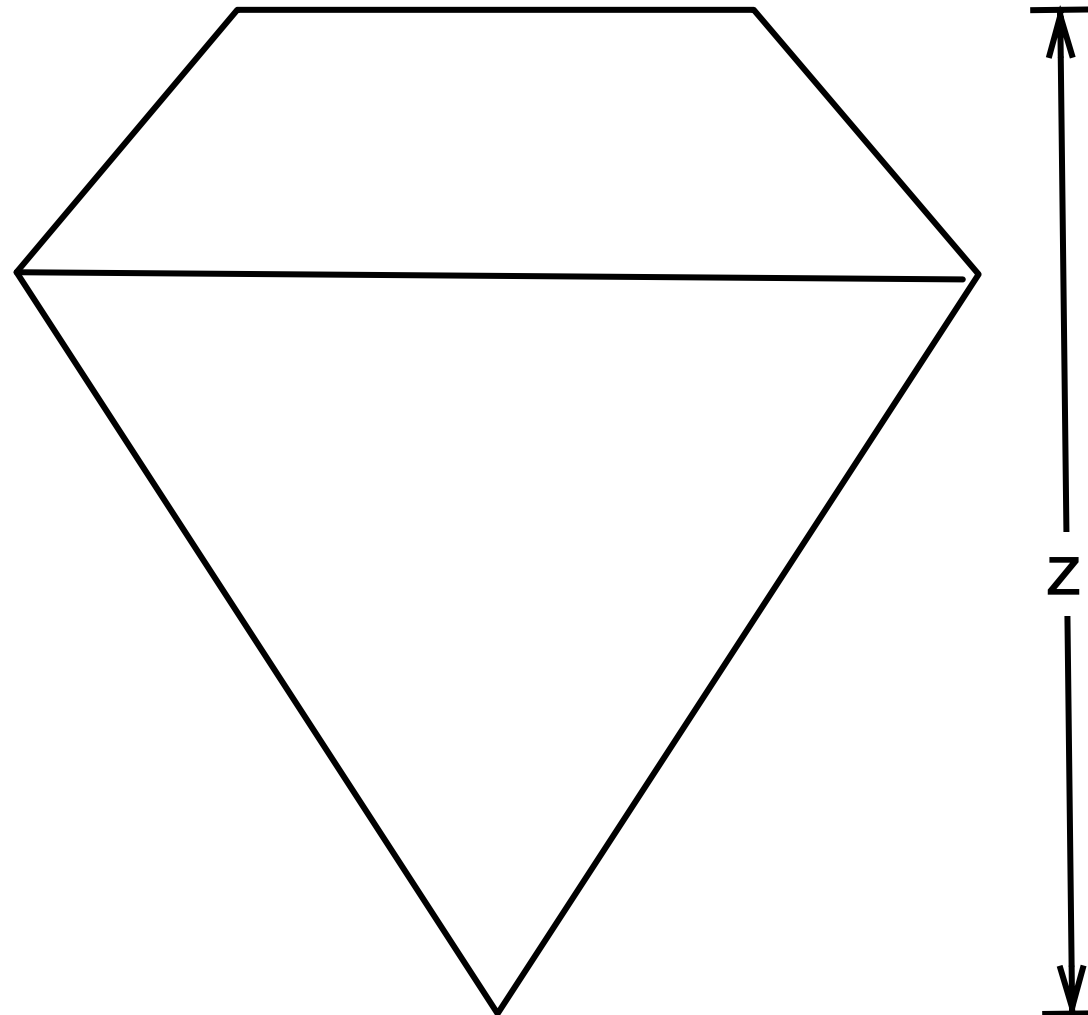
# Diamonds data

- ~**54,000** round diamonds from <http://www.diamondse.info/>
- Carat, color, clarity, cut
- Total depth, table, depth, width, height
- Price



← x →

← table width →



$$\text{depth} = z / \text{diameter}$$
$$\text{table} = \text{table width} / x * 100$$

COLOR GRADING SCALE																						
																						
D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z
Colorless			Near Colorless				Faint Yellow			Very Light Yellow				Light Yellow								



IF



VVS1



VVS2



VSI



Illustration of inclusions as seen under X10 magnification

VS2



SII



SI2



I1



# **Histogram & bar charts**

# Your turn

What types of plots do the following lines of code return?

```
qplot(x, z, data = diamonds)
```

```
qplot(x, data = diamonds)
```

```
qplot(cut, data = diamonds)
```

# Default geoms for qplot

Two variables → scatterplot (point)

One continuous variable → histogram

One categorical variable → bar chart

# Parameters

Similar to aesthetics.

A parameter is input that controls the appearance of the graph, *but does not map appearance to data.*

e.g. binwidths in a histogram

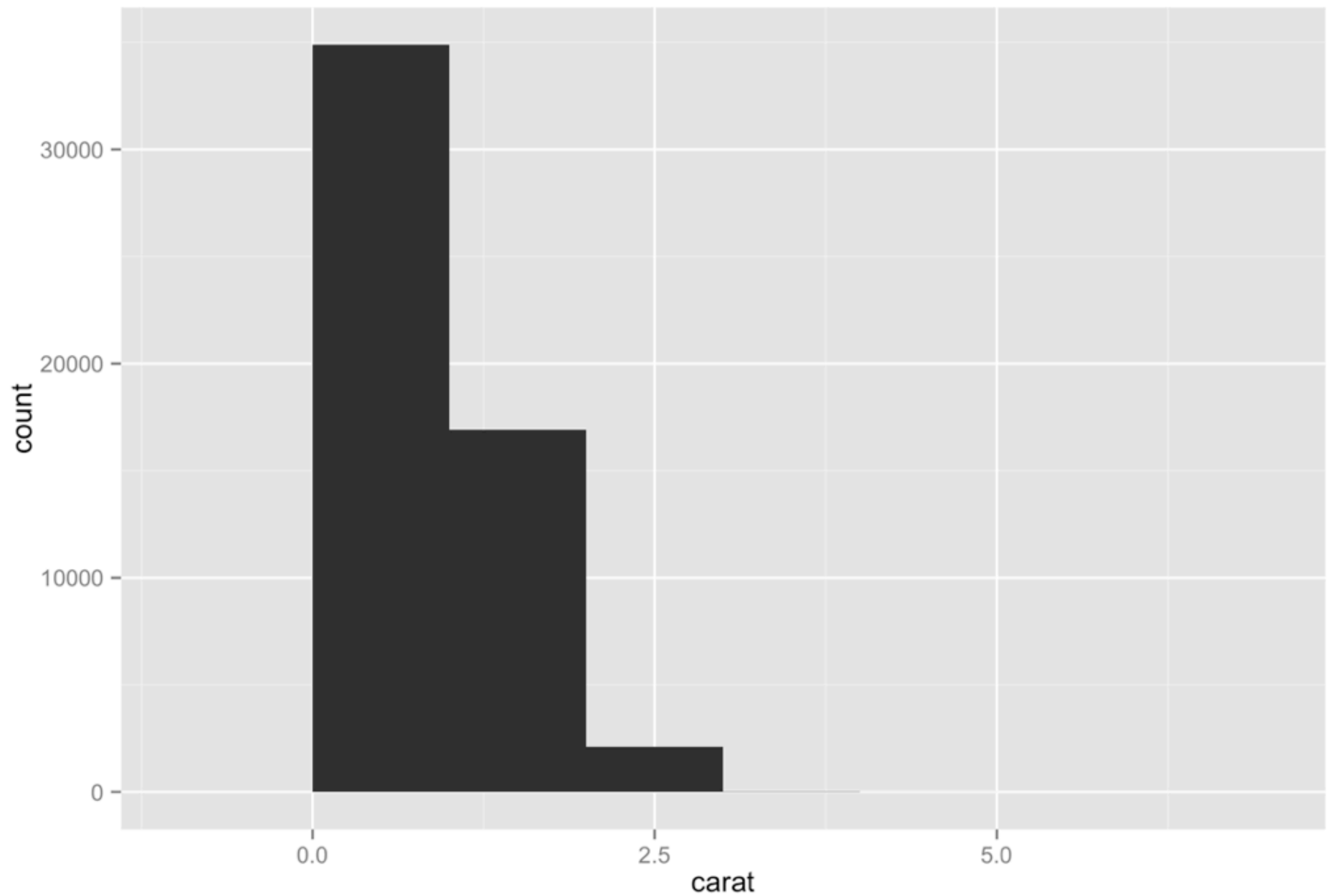
# Parameters

parameter  
name

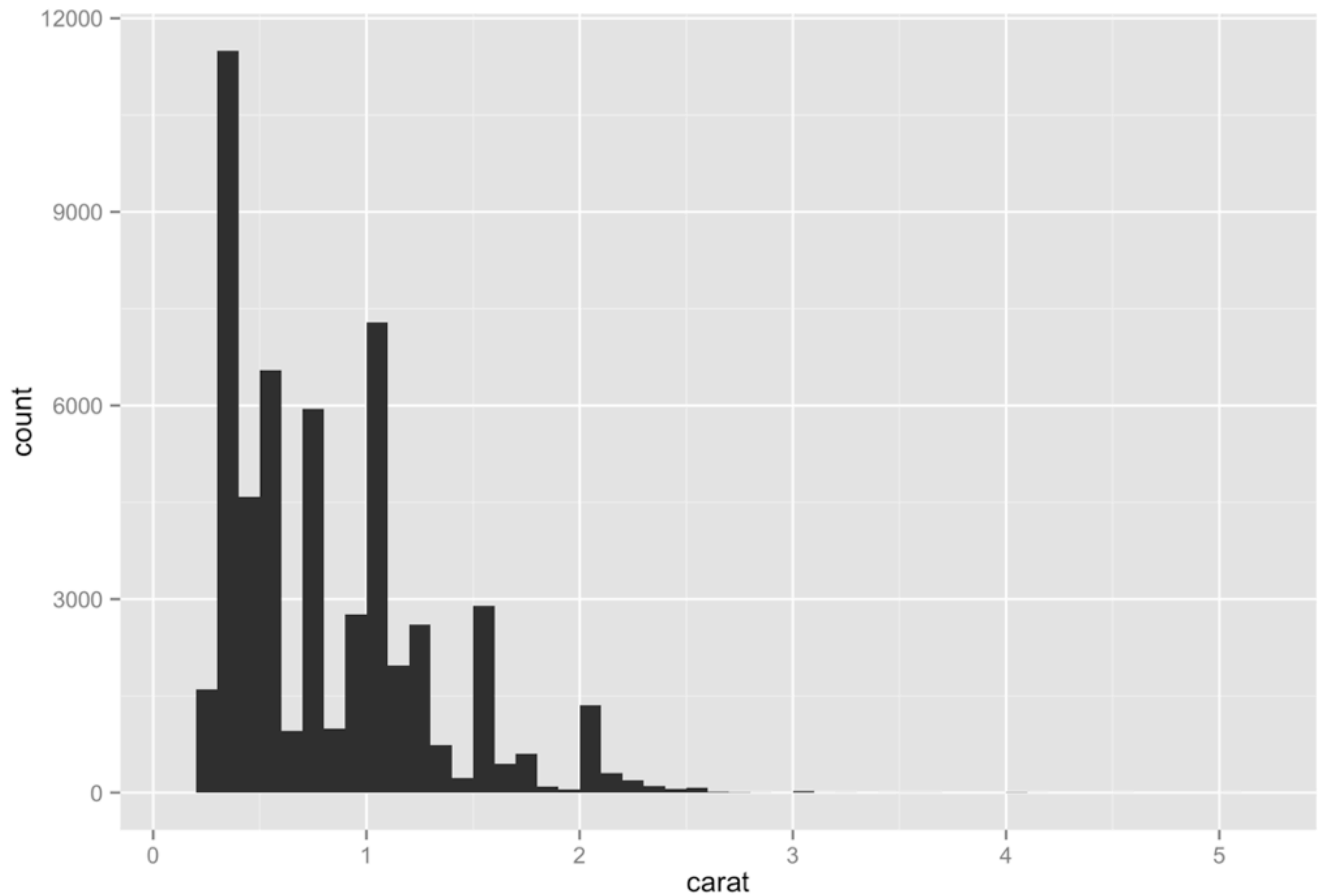
value

```
qplot(displ, data = mpg, binwidth = 1)
```

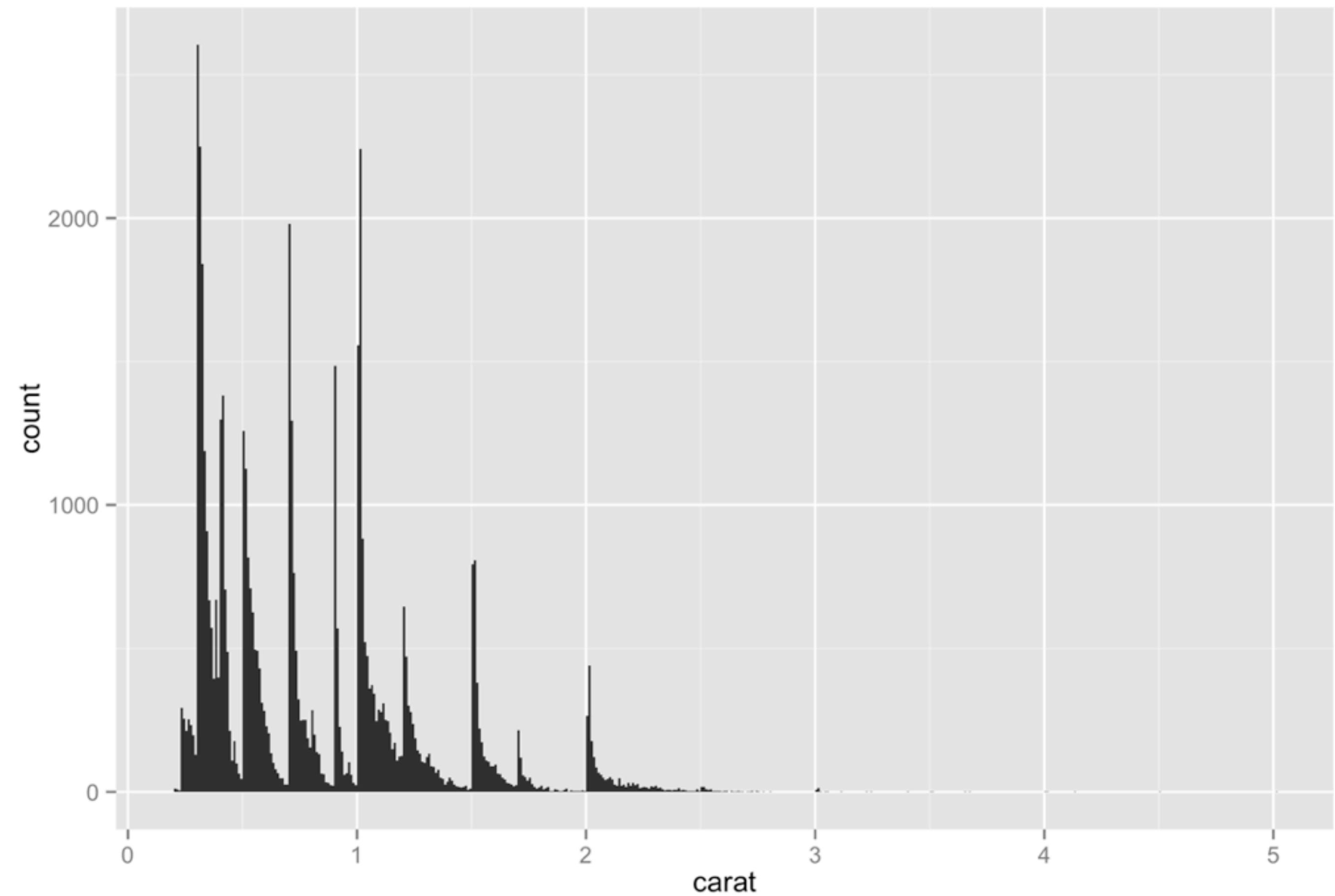




```
qplot(carat, data = diamonds, binwidth = 1)
```

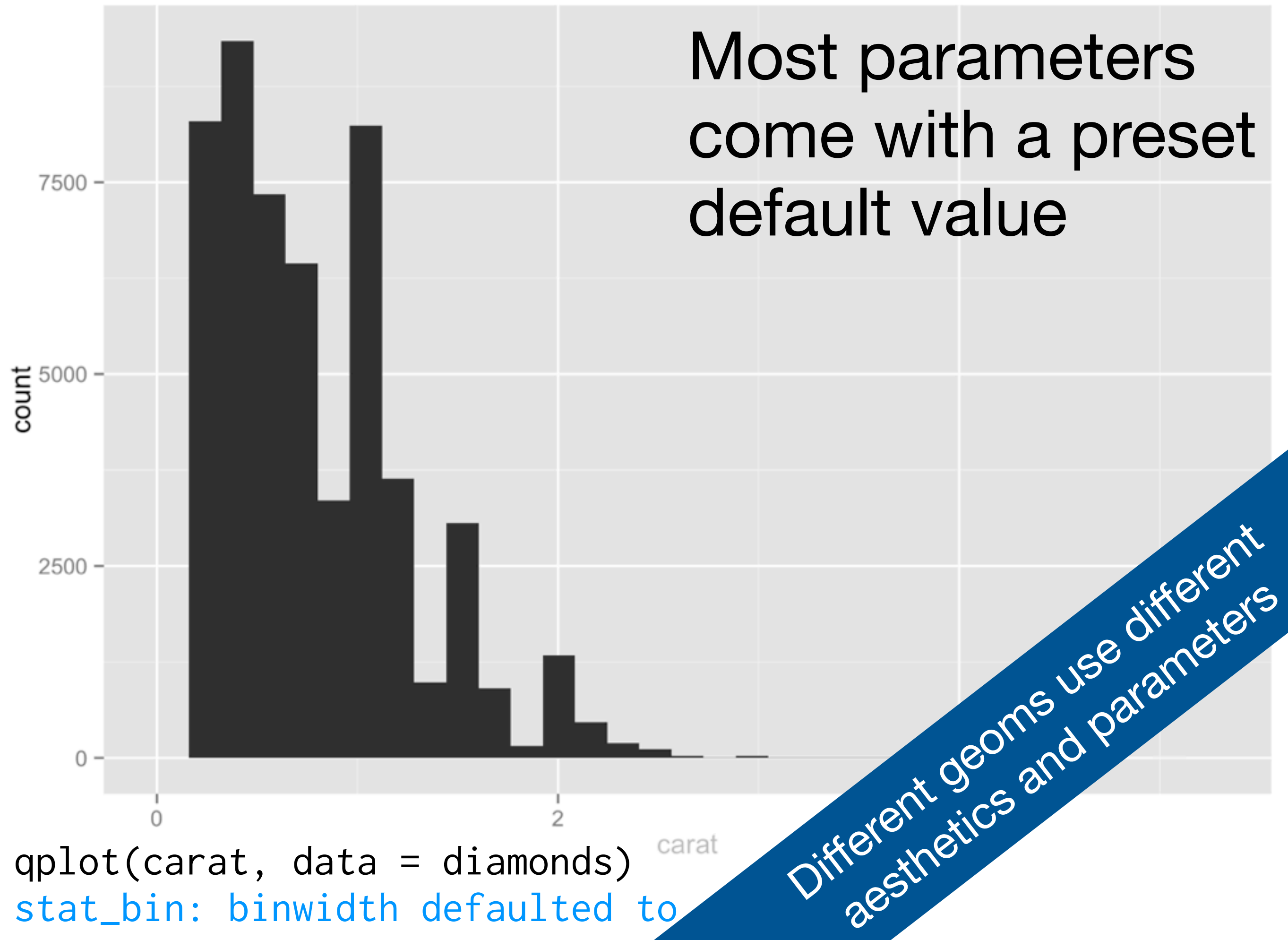


```
qplot(carat, data = diamonds, binwidth = 0.1)
```



```
qplot(carat, data = diamonds, binwidth = 0.01)
```

Most parameters  
come with a preset  
default value



```
qplot(carat, data = diamonds)  
stat_bin: binwidth defaulted to
```

# Your turn

Examine the distribution of price at different binwidths.

Do you spot anything odd?

```
qplot(price, data = diamonds)
```

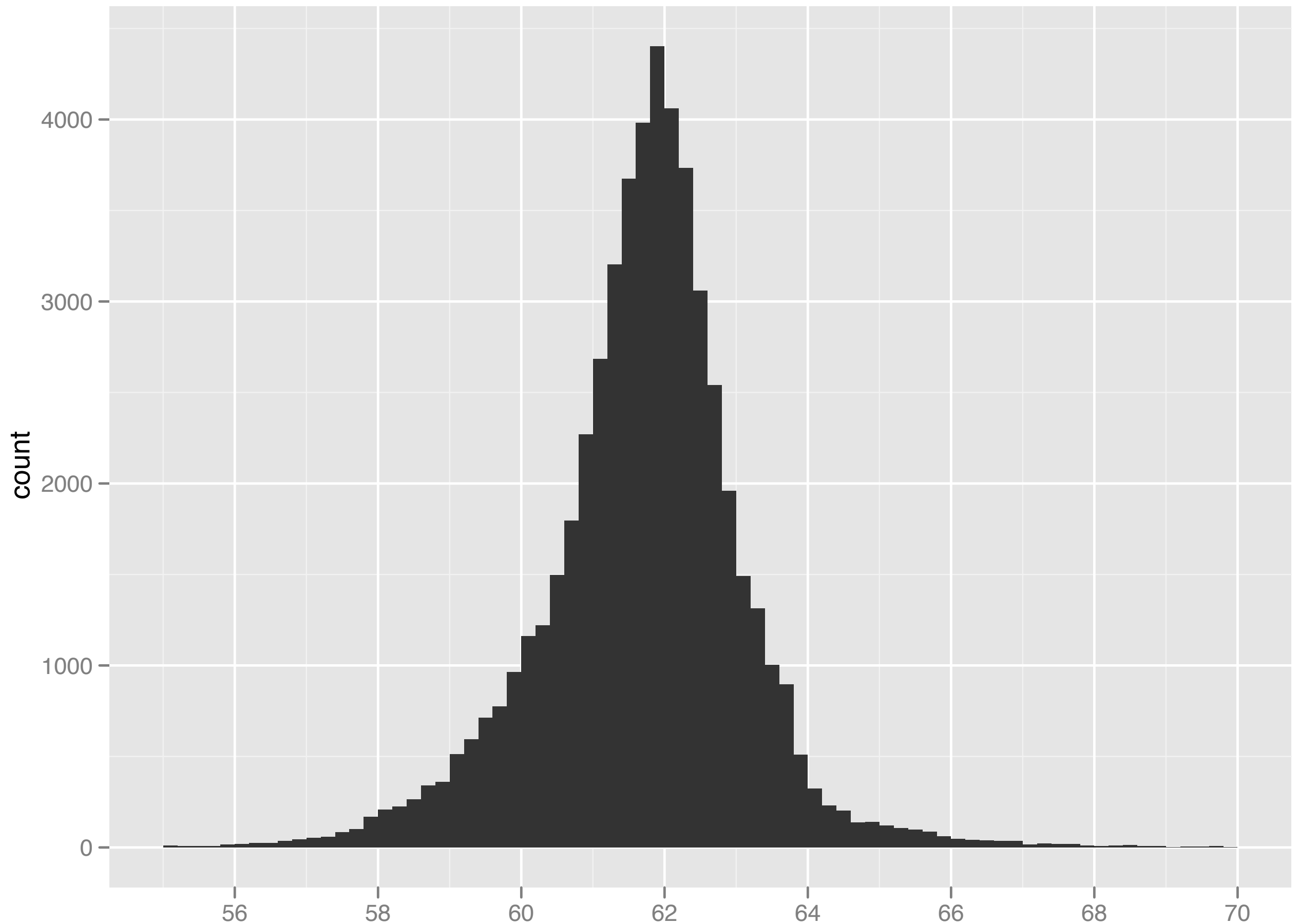
Hint:  $0 < \text{price} < 18823$ .  
Do not set `binwidth = 1`!

```
qplot(price, data = diamonds)
qplot(price, data = diamonds, binwidth = 500)
qplot(price, data = diamonds, binwidth = 100)
qplot(price, data = diamonds, binwidth = 50)
```

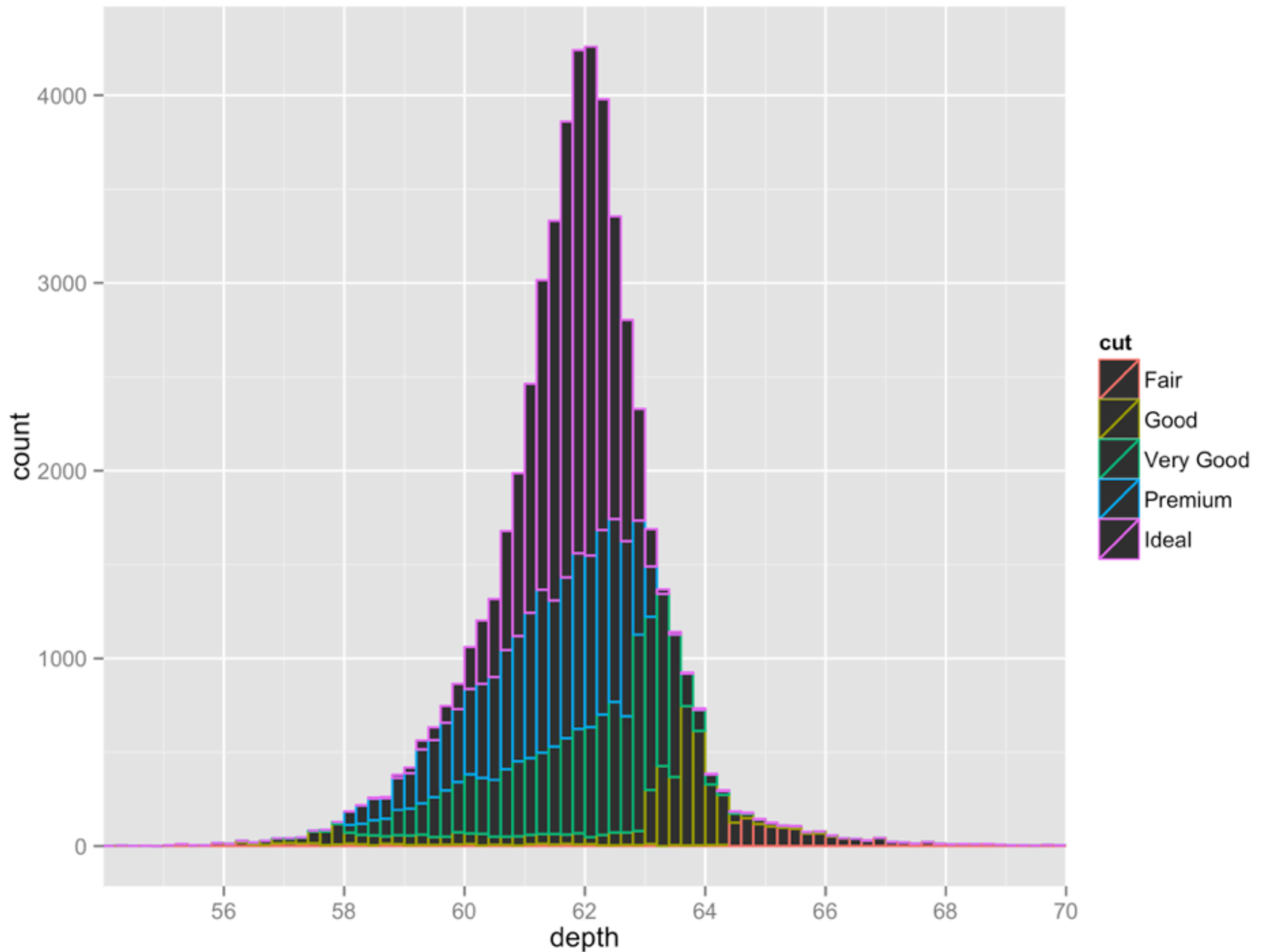


# Additional variables

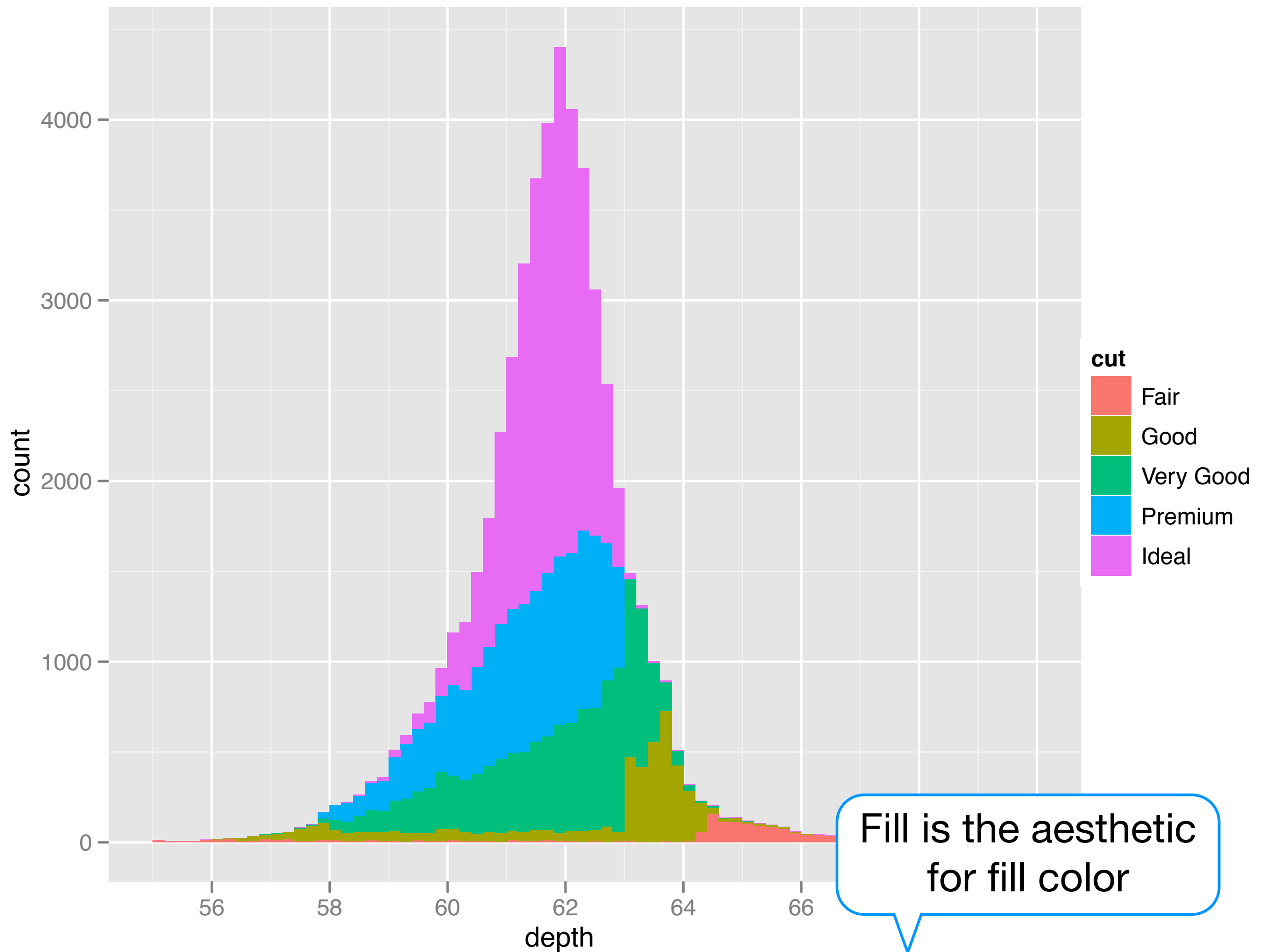
Often switching geoms is more effective than adding aesthetics or faceting to a histogram



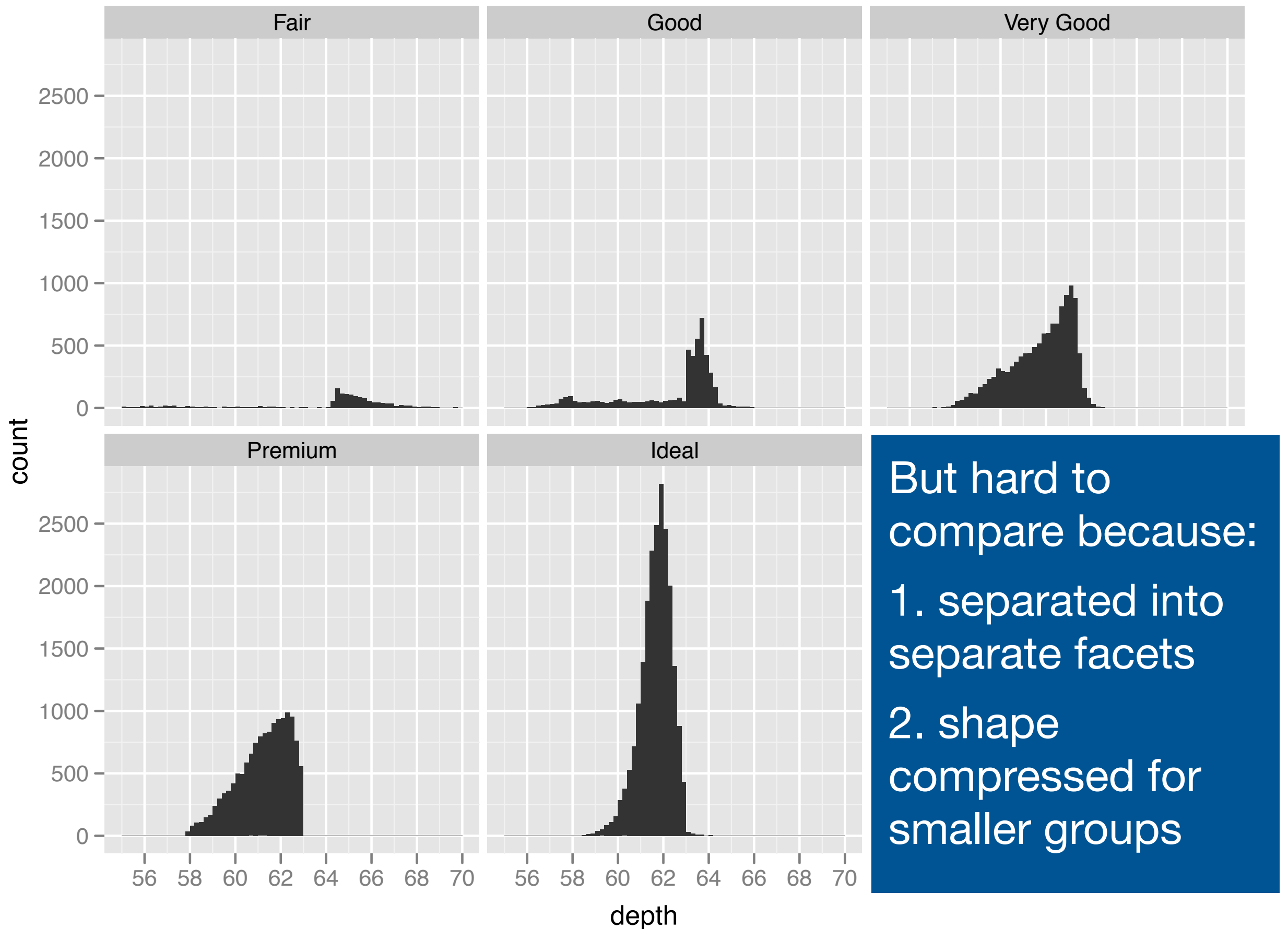
```
qplot(depth, data = diamonds, binwidth = 0.2)
```



```
qplot(depth, data = diamonds, binwidth = 0.2, color = cut)
```

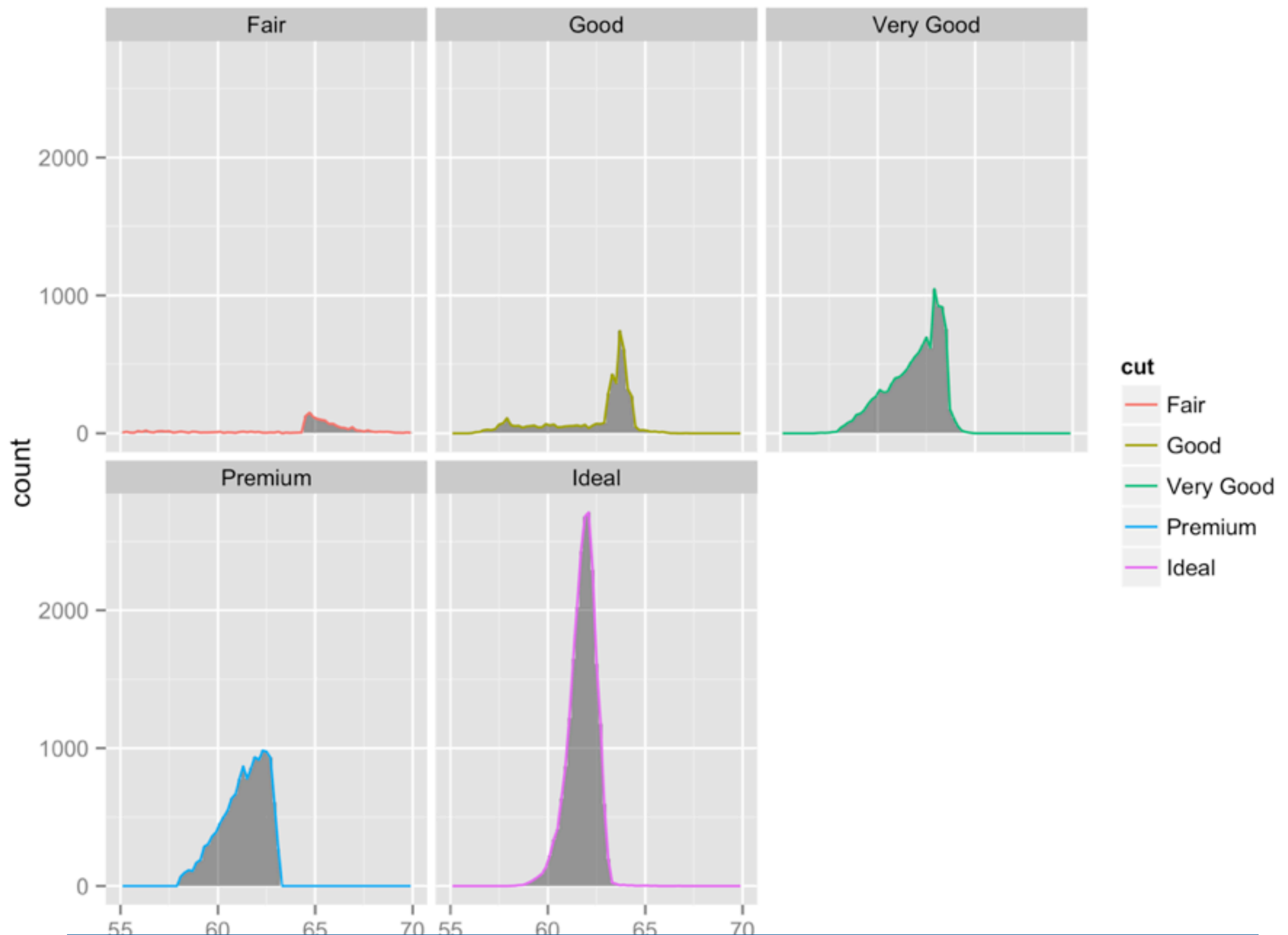


```
qplot(depth, data = diamonds, binwidth = 0.2, fill = cut)
```

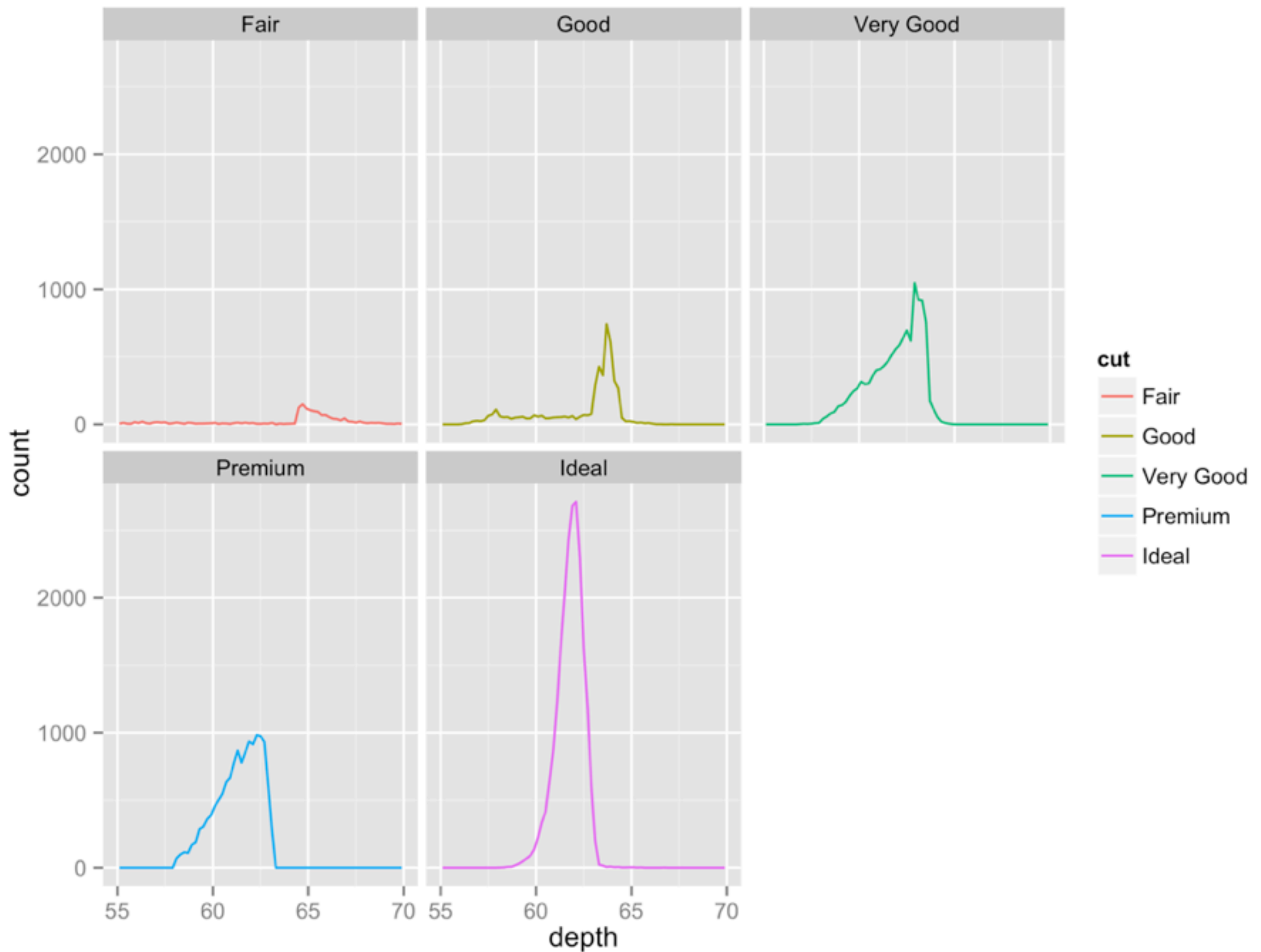


But hard to  
compare because:  
1. separated into  
separate facets  
2. shape  
compressed for  
smaller groups

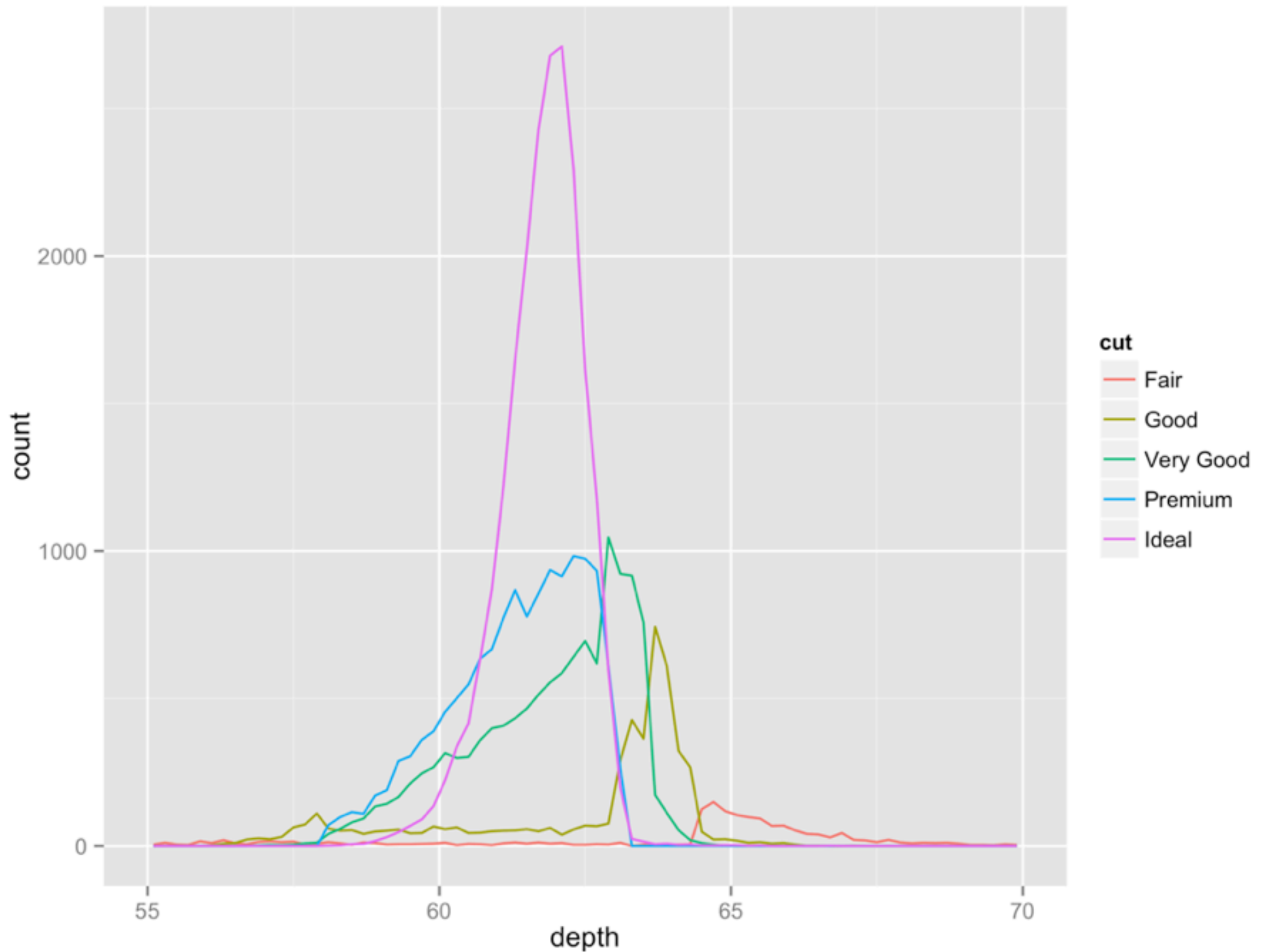
```
qplot(depth, data = diamonds, binwidth = 0.2) +  
  facet_wrap(~ cut)
```



What if we just drew a line along the tops of the histograms, and threw away the bars?

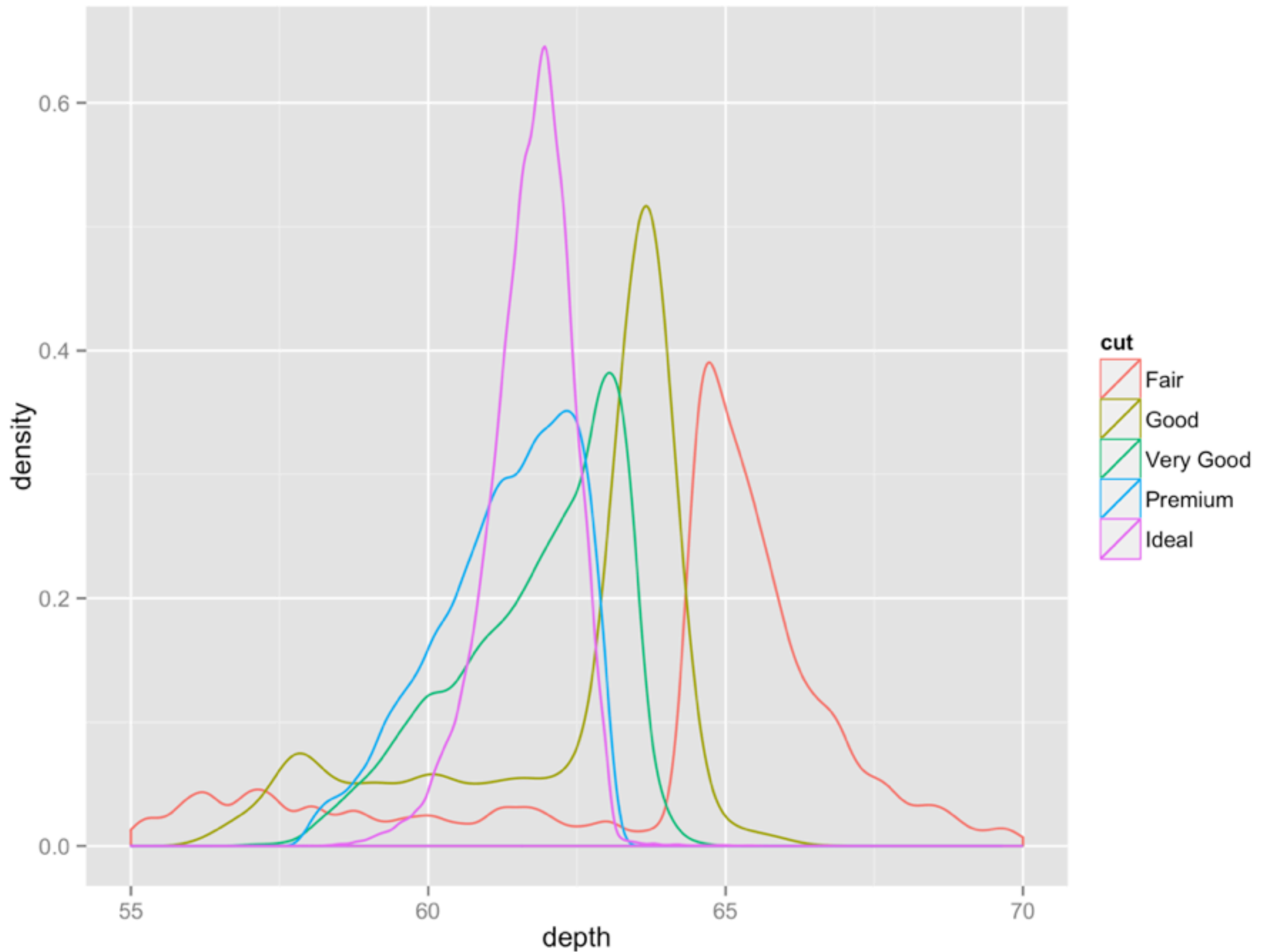


```
qplot(depth, data = diamonds, geom = "freqpoly", color = cut,  
       binwidth = 0.2) + facet_wrap(~ cut)
```



```
qplot(depth, data = diamonds, geom = "freqpoly",  
      color = cut, binwidth = 0.2)
```

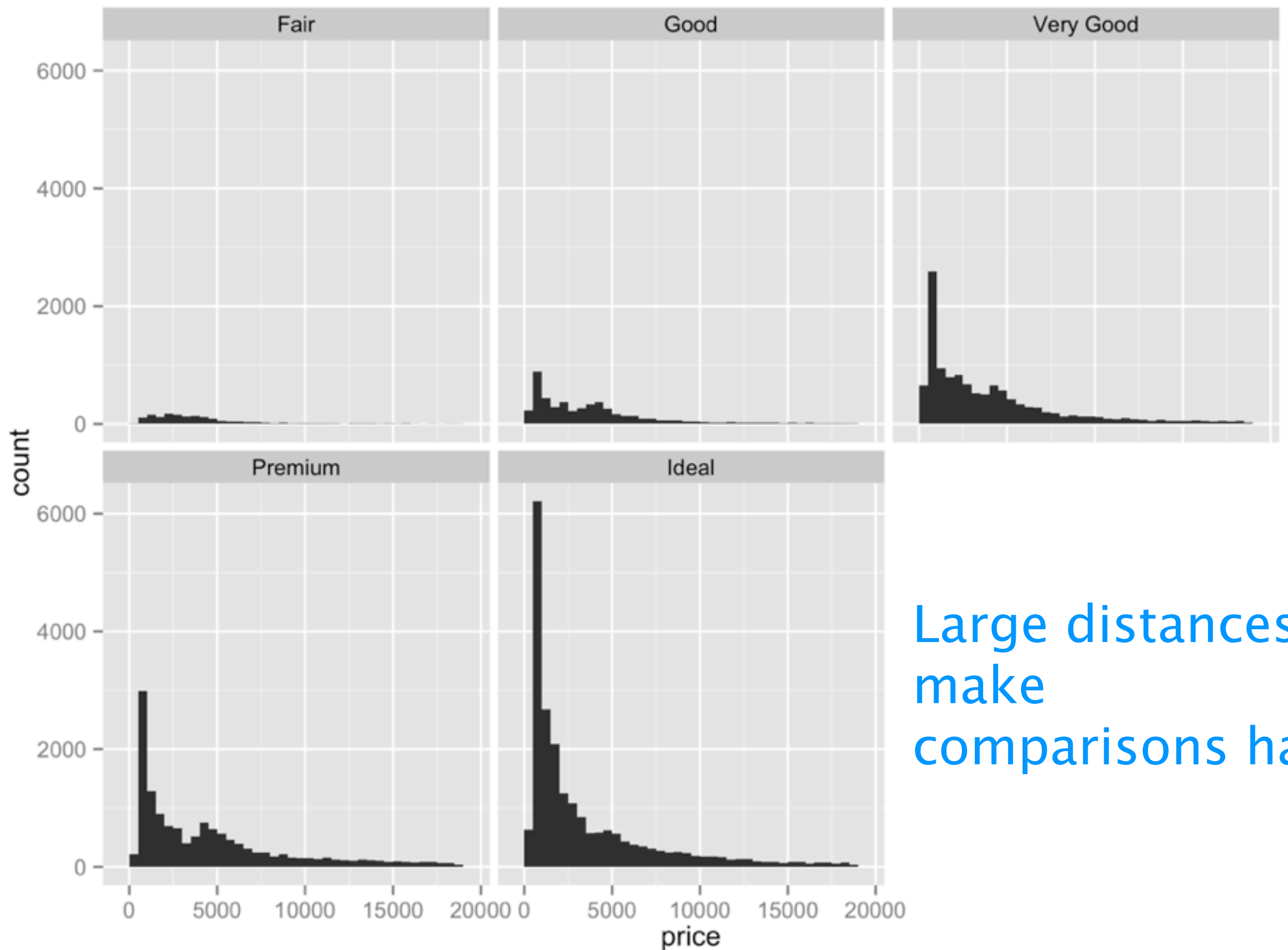




```
qplot(depth, data = diamonds, geom = "density",  
color = cut)
```

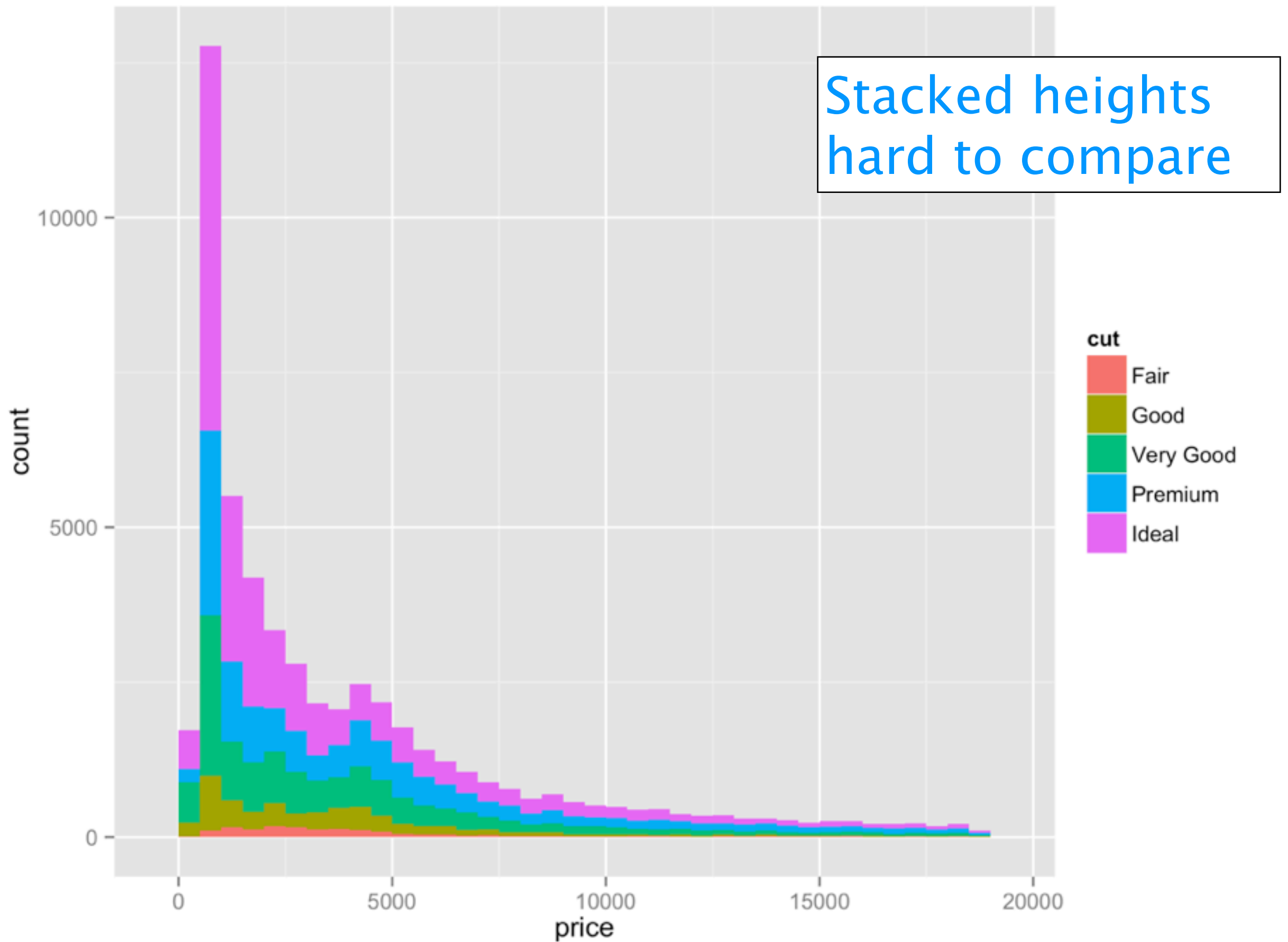
# Your turn

Compare the distribution of price for the different cuts. Does anything seem unusual?

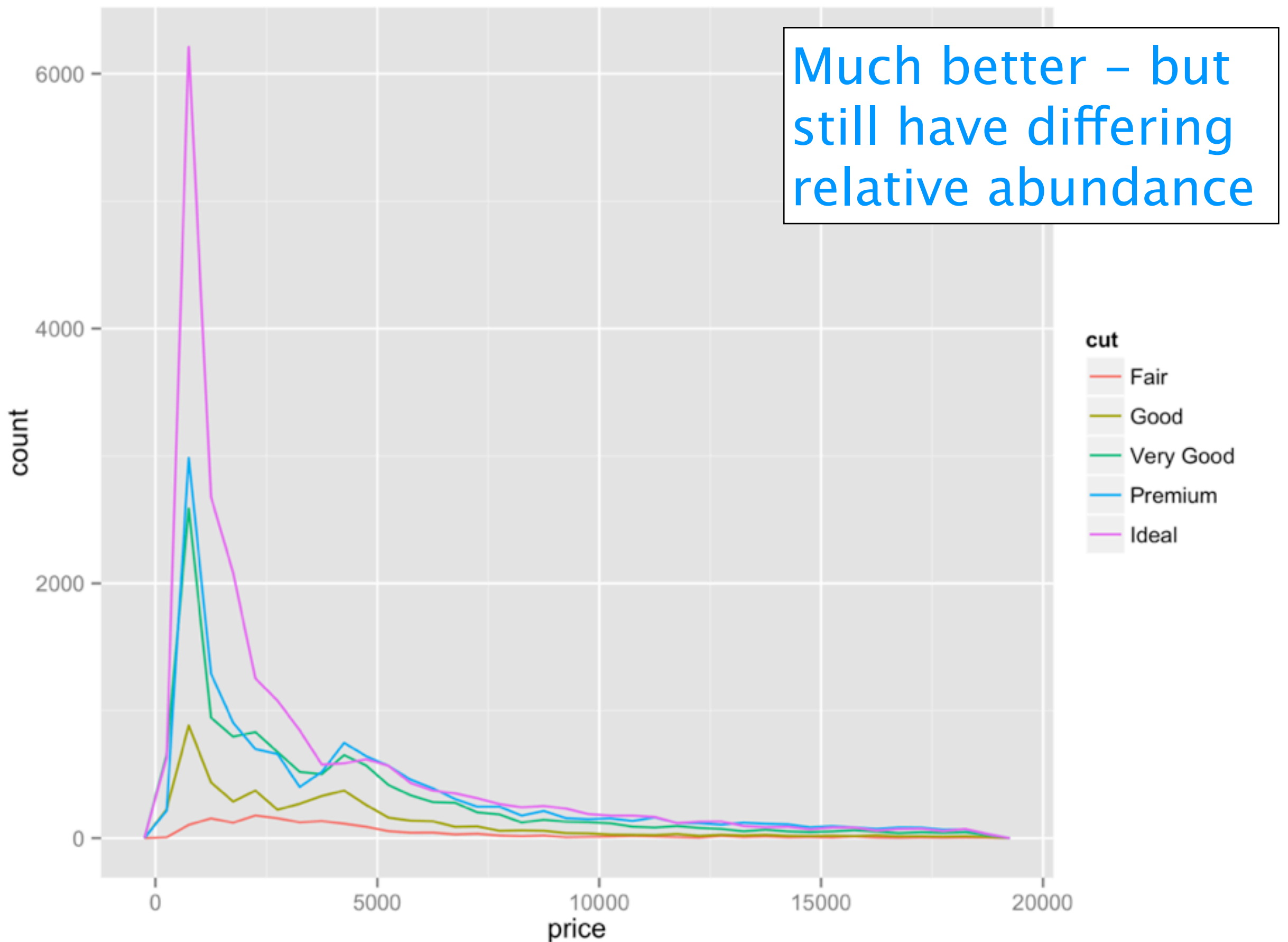


Large distances  
make  
comparisons hard

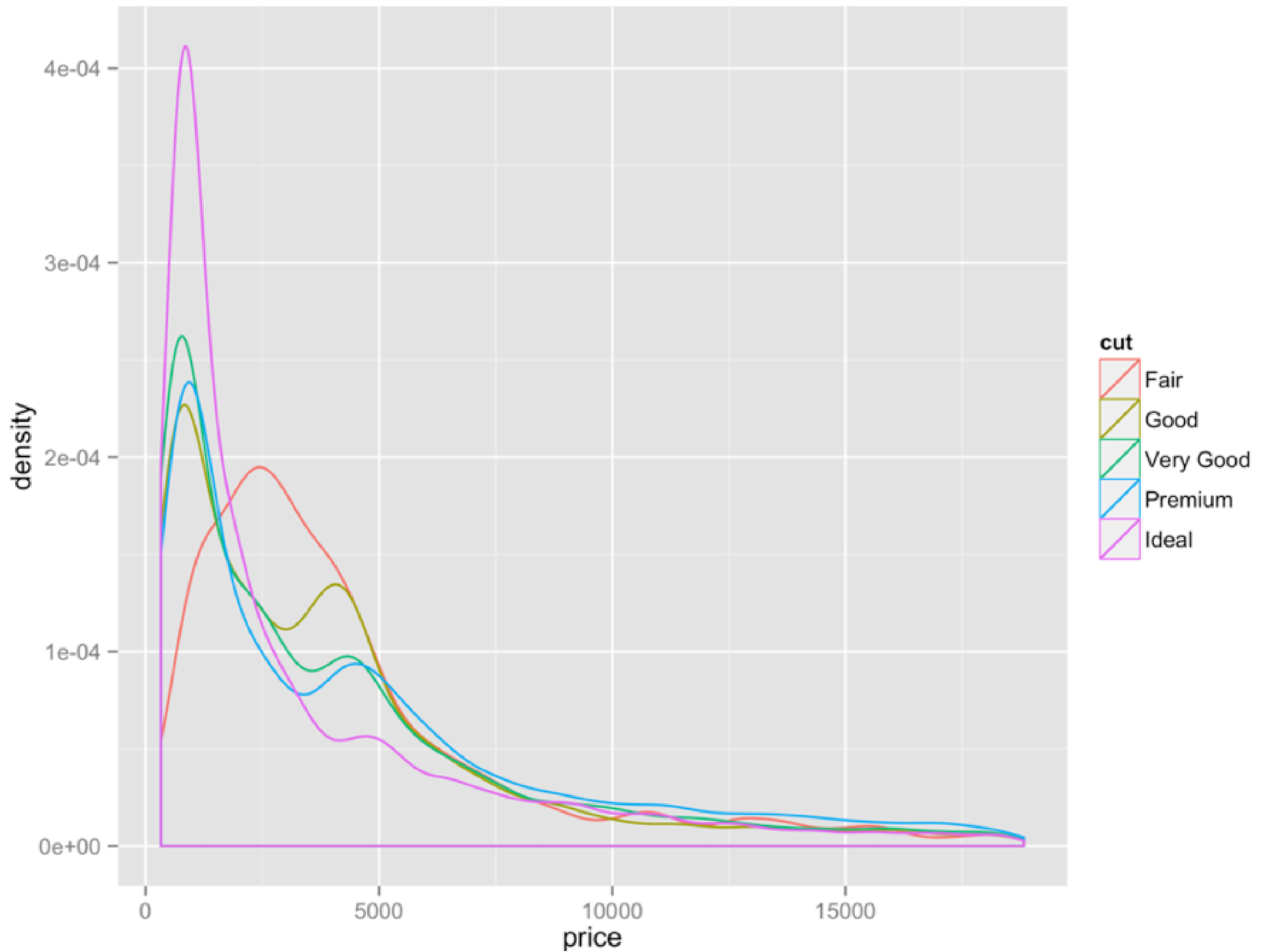
```
qplot(price, data = diamonds, binwidth = 500) +  
  facet_wrap(~ cut)
```



```
qplot(price, data = diamonds, binwidth = 500,  
      fill = cut)
```

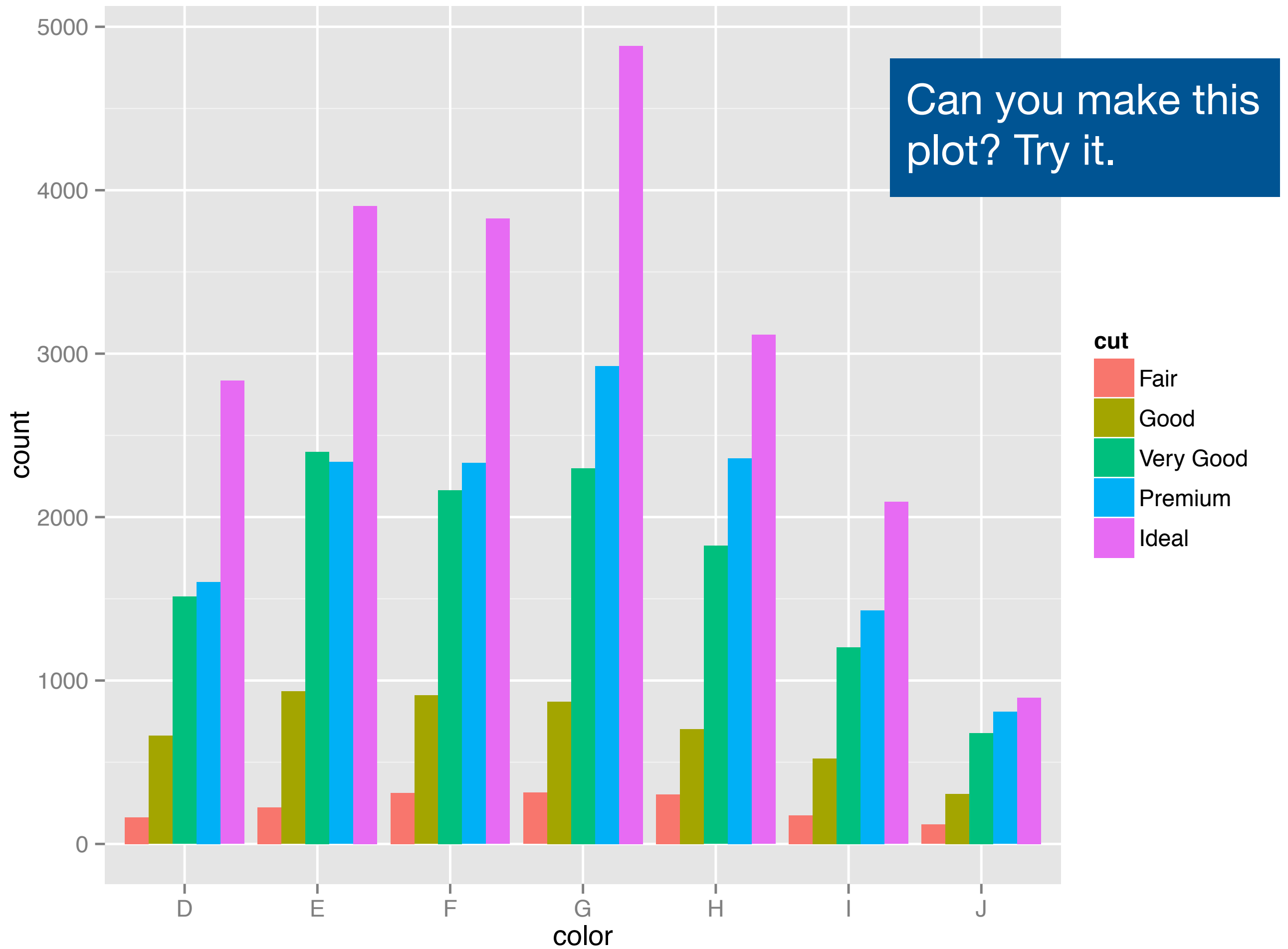


```
qplot(price, data = diamonds, binwidth = 500,  
      geom = "freqpoly", color = cut)
```



```
qplot(price, data = diamonds, geom = "density",  
color = cut)
```

# **Position adjustments**





# Your turn

What do each of the position adjustments below do?

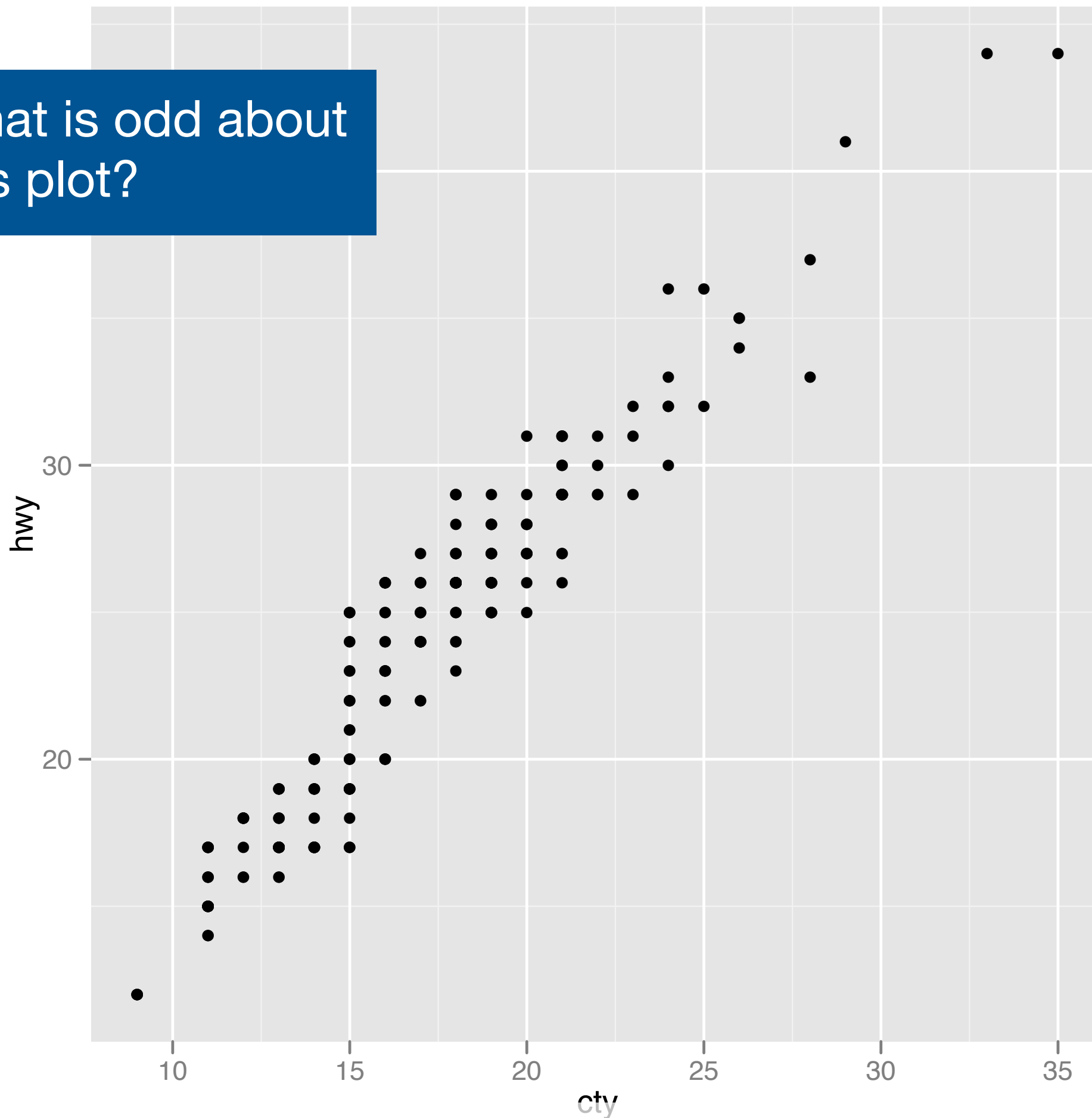
```
qplot(color, data = diamonds, fill = cut,  
position = "stack")
```

```
qplot(color, data = diamonds, fill = cut,  
position = "dodge")
```

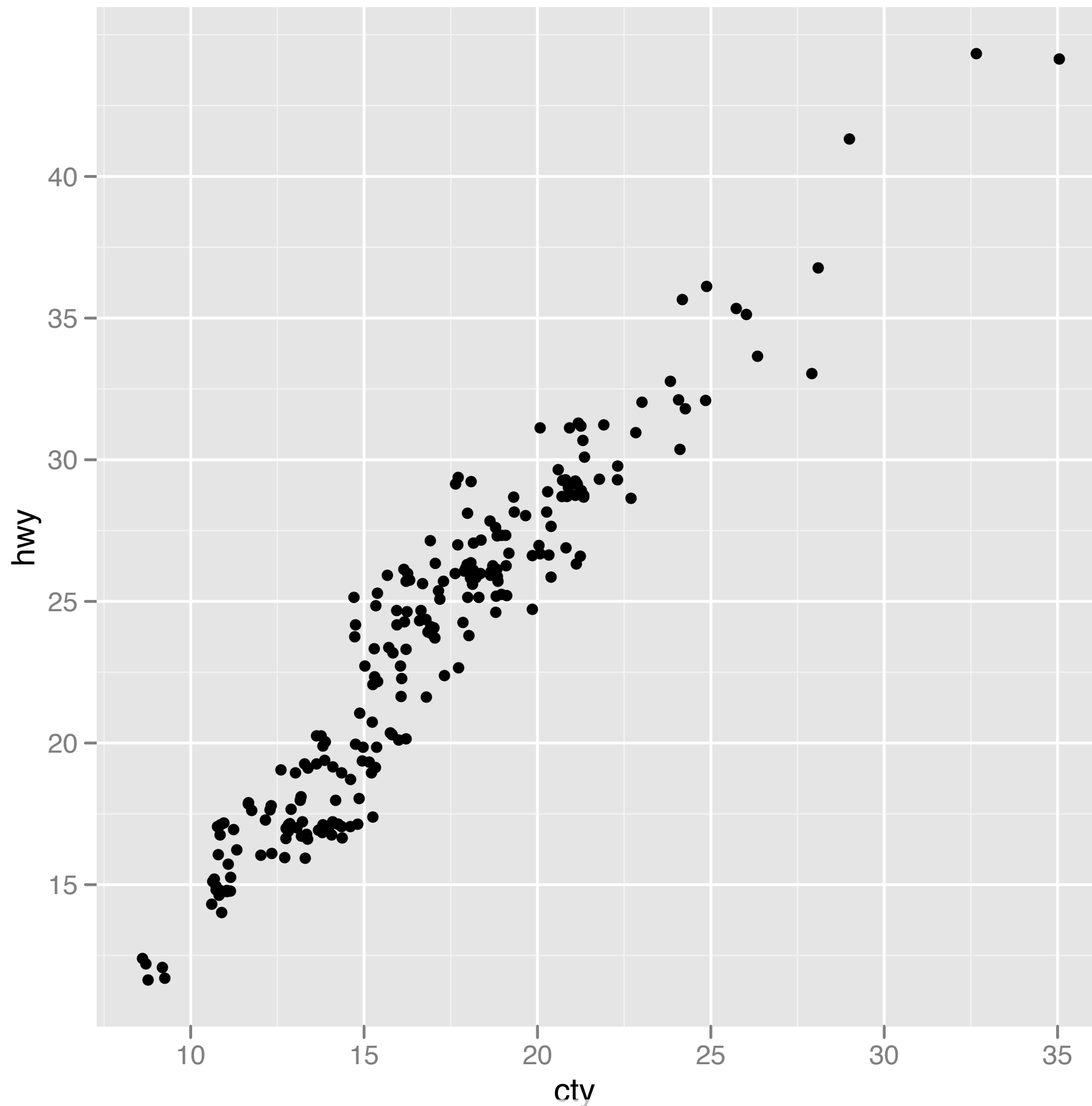
```
qplot(color, data = diamonds, fill = cut,  
position = "identity")
```

```
qplot(color, data = diamonds, fill = cut,  
position = "fill")
```

What is odd about this plot?



```
qplot(cty, hwy, data = mpg)
```



```
qplot(cty, hwy, data = mpg, position = "jitter")
```

Position adjustment	effect
identity	no adjustment
stack	colliding objects plotted <i>above</i> each other
dodge	colliding objects plotted <i>beside</i> each other
fill	available space divided proportionately
jitter	random noise added to placement of each object

# **Grammar of graphics**

# Summary

qplot + aesthetics = 10's of plots

qplot + geoms = 100's of plots

qplot + geoms + aesthetics = 1000's of plots

qplot + geoms + aesthetics + position adj = 100,000's



“If any number of magnitudes are each the same multiple of the same number of other magnitudes, then the sum is that multiple of the sum.”

*Euclid, ~300 BC*





“If any number of magnitudes are each the same multiple of the same number of other magnitudes, then the sum is that multiple of the sum.”

*Euclid, ~300 BC*

$$m \sum x_i = \sum mx_i$$

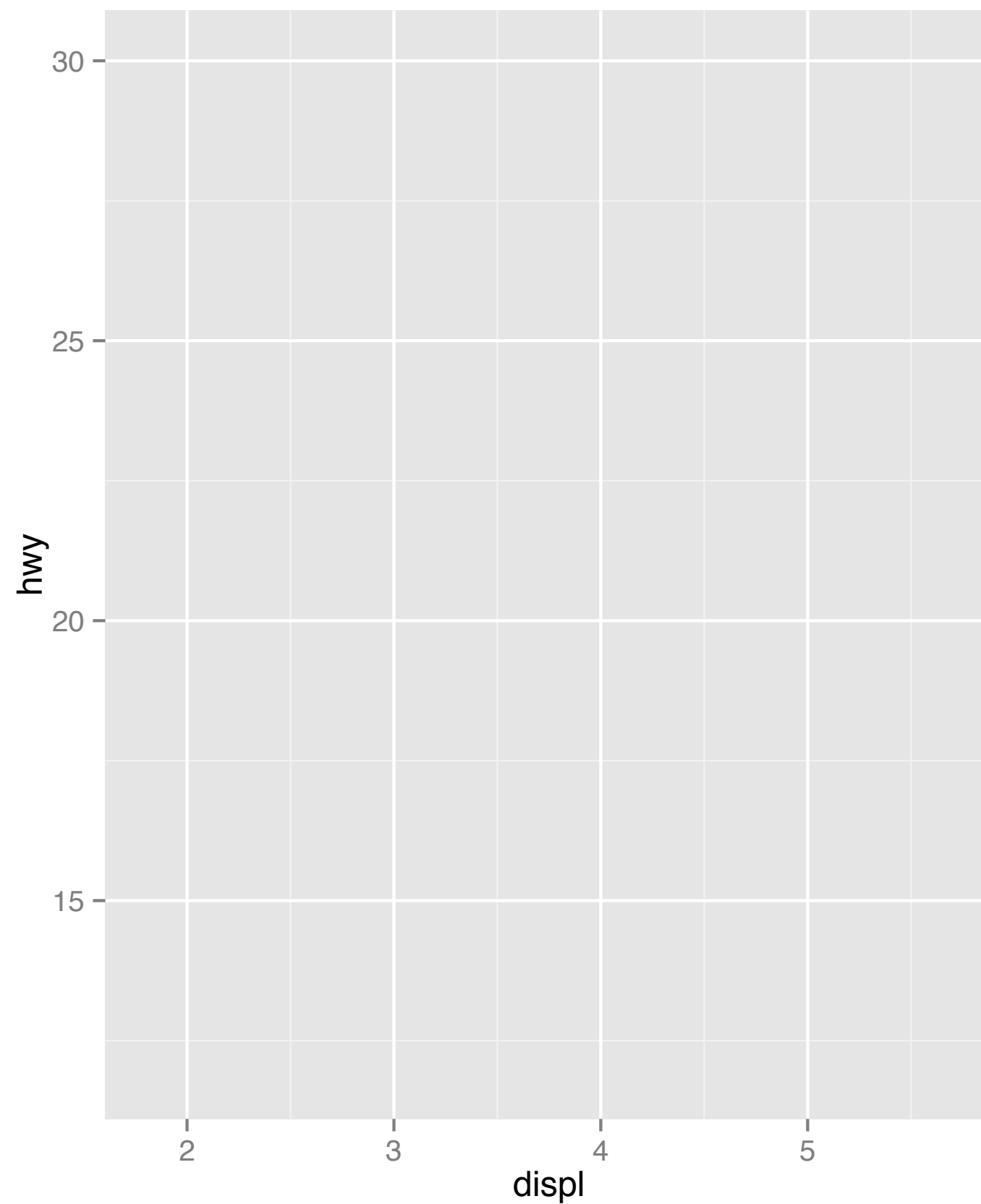


# The grammar of graphics

An abstraction which makes it easier to describe, make, and reason about graphics.

Developed by Leland Wilkinson, particularly in “*The Grammar of Graphics*” 1999/2005

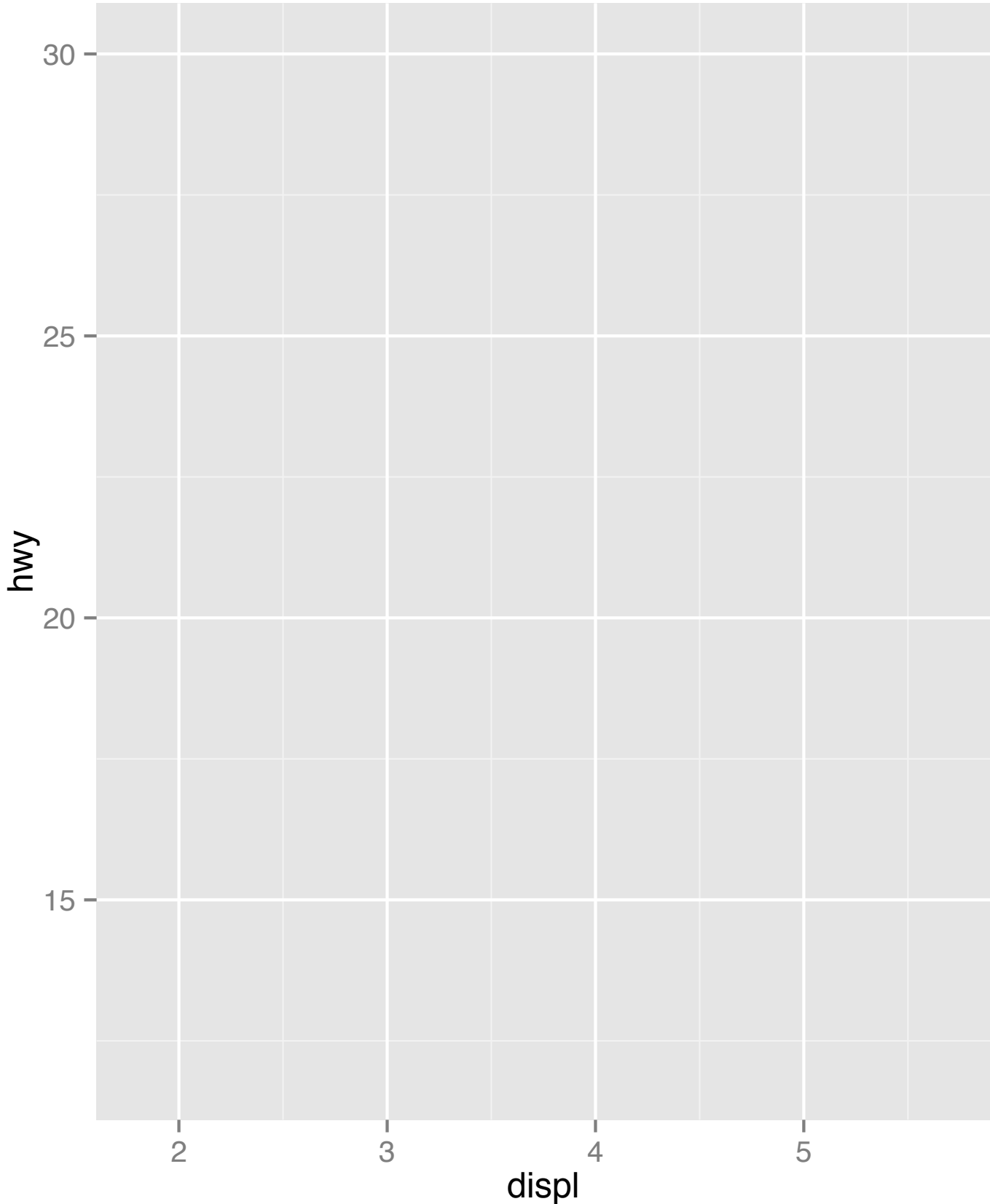
How to build a plot



Coordinate system

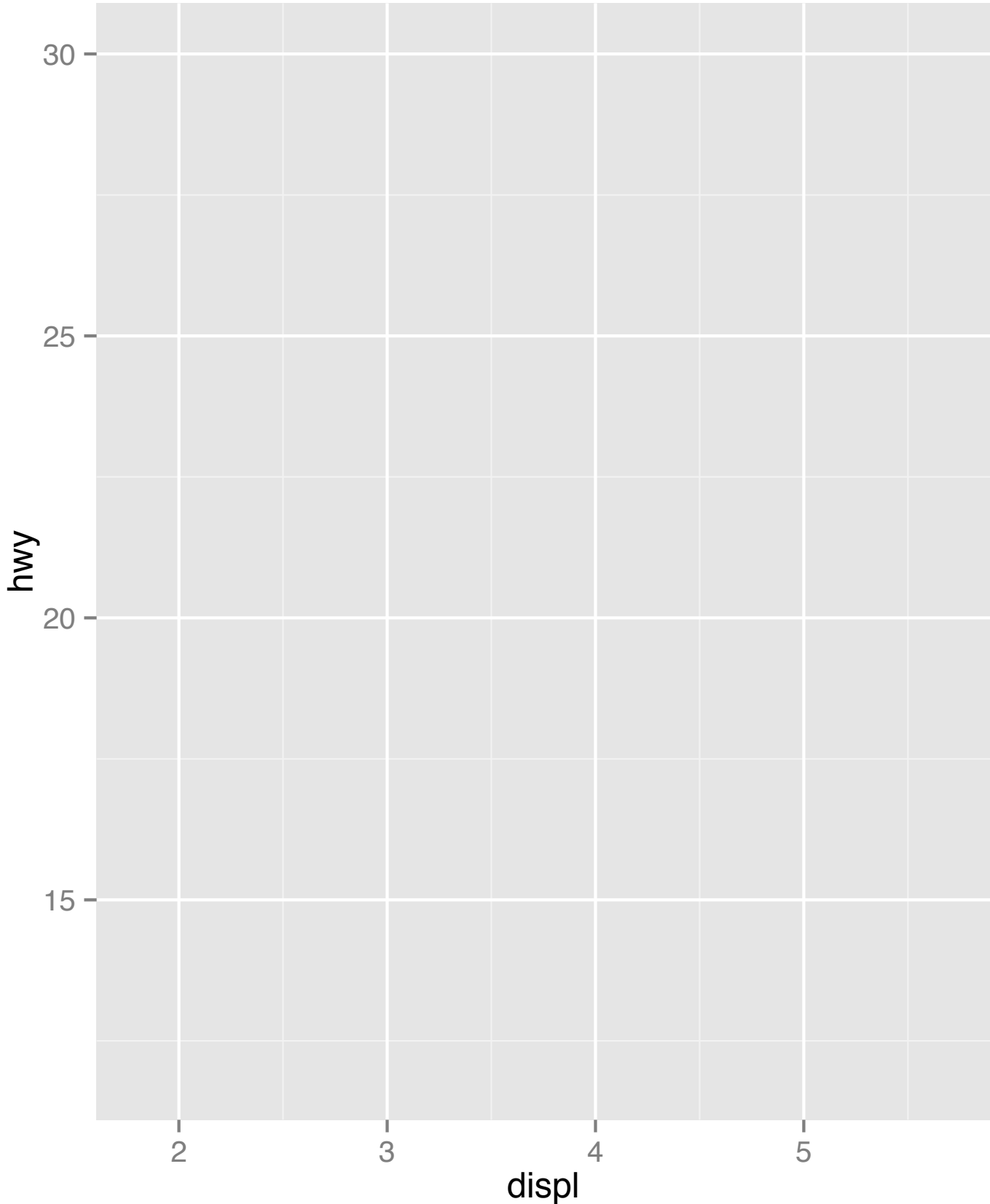
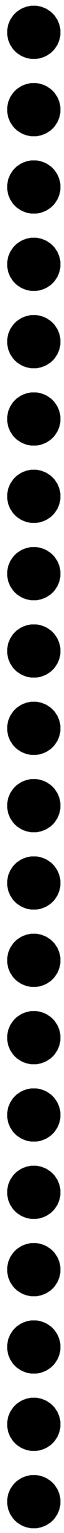
hwy	displ	cyl	class
17	5.0	8	suv
20	2.7	4	pickup
17	4.0	6	suv
25	2.8	6	compact
27	3.1	6	compact
30	2.0	4	compact
25	2.8	6	compact
23	2.8	6	compact
26	3.0	6	midsize
17	5.4	8	pickup
28	2.5	5	subcompact
29	3.5	6	midsize
26	2.4	4	midsize
29	2.0	4	midsize
15	5.4	8	pickup
29	1.8	4	compact
18	5.7	8	suv
12	4.7	8	pickup
26	2.8	6	compact
24	3.3	6	minivan

Data



Coordinate system

hwy	displ	cyl	class
17	5.0	8	suv
20	2.7	4	pickup
17	4.0	6	suv
25	2.8	6	compact
27	3.1	6	compact
30	2.0	4	compact
25	2.8	6	compact
23	2.8	6	compact
26	3.0	6	midsize
17	5.4	8	pickup
28	2.5	5	subcompact
29	3.5	6	midsize
26	2.4	4	midsize
29	2.0	4	midsize
15	5.4	8	pickup
29	1.8	4	compact
18	5.7	8	suv
12	4.7	8	pickup
26	2.8	6	compact
24	3.3	6	minivan



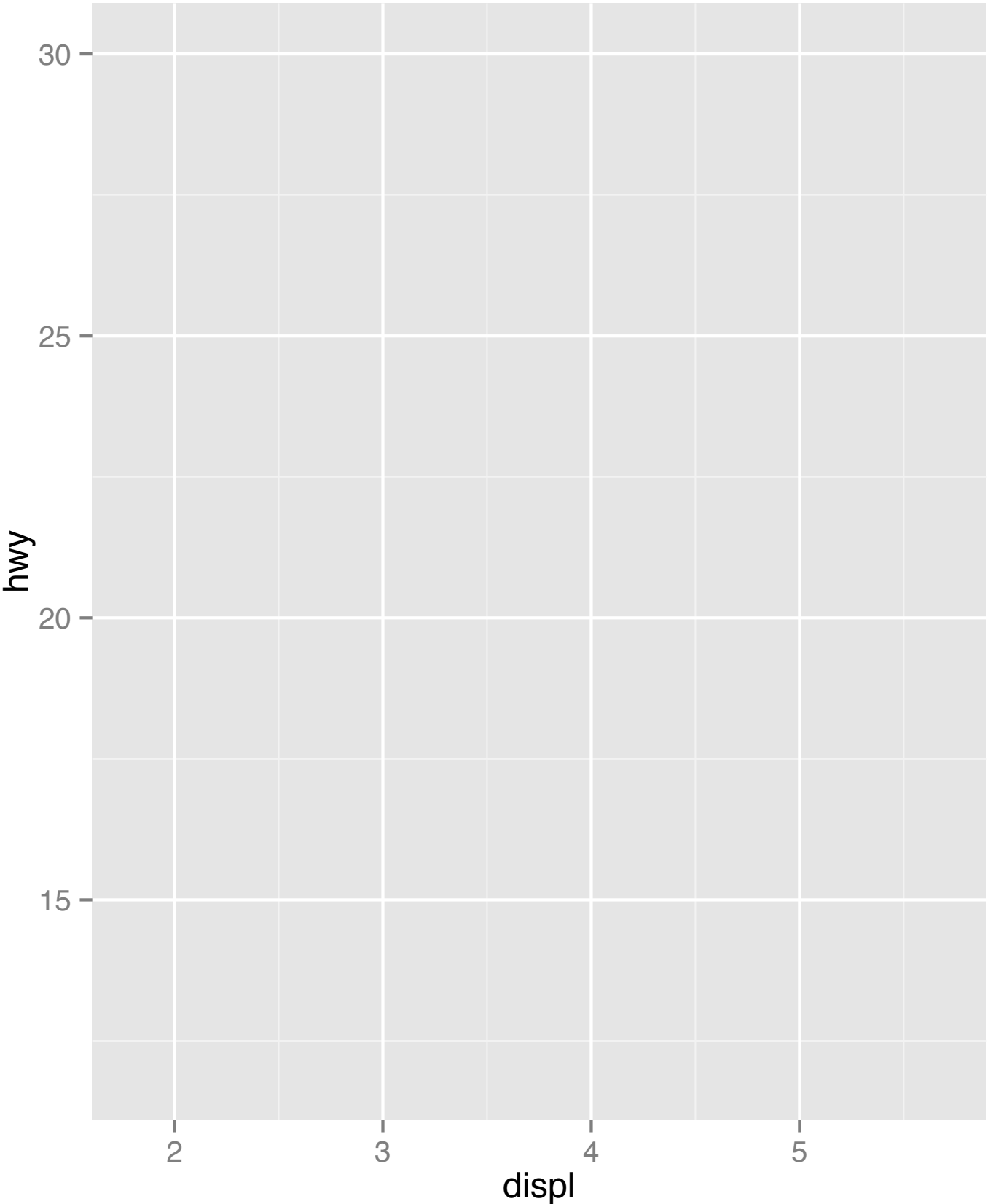
Data

Geom

Coordinate system

# Aesthetic mappings

hwy	displ	cyl	class
17	5.0	8	suv
20	2.7	4	pickup
17	4.0	6	suv
25	2.8	6	compact
27	3.1	6	compact
30	2.0	4	compact
25	2.8	6	compact
23	2.8	6	compact
26	3.0	6	midsize
17	5.4	8	pickup
28	2.5	5	subcompact
29	3.5	6	midsize
26	2.4	4	midsize
29	2.0	4	midsize
15	5.4	8	pickup
29	1.8	4	compact
18	5.7	8	suv
12	4.7	8	pickup
26	2.8	6	compact
24	3.3	6	minivan



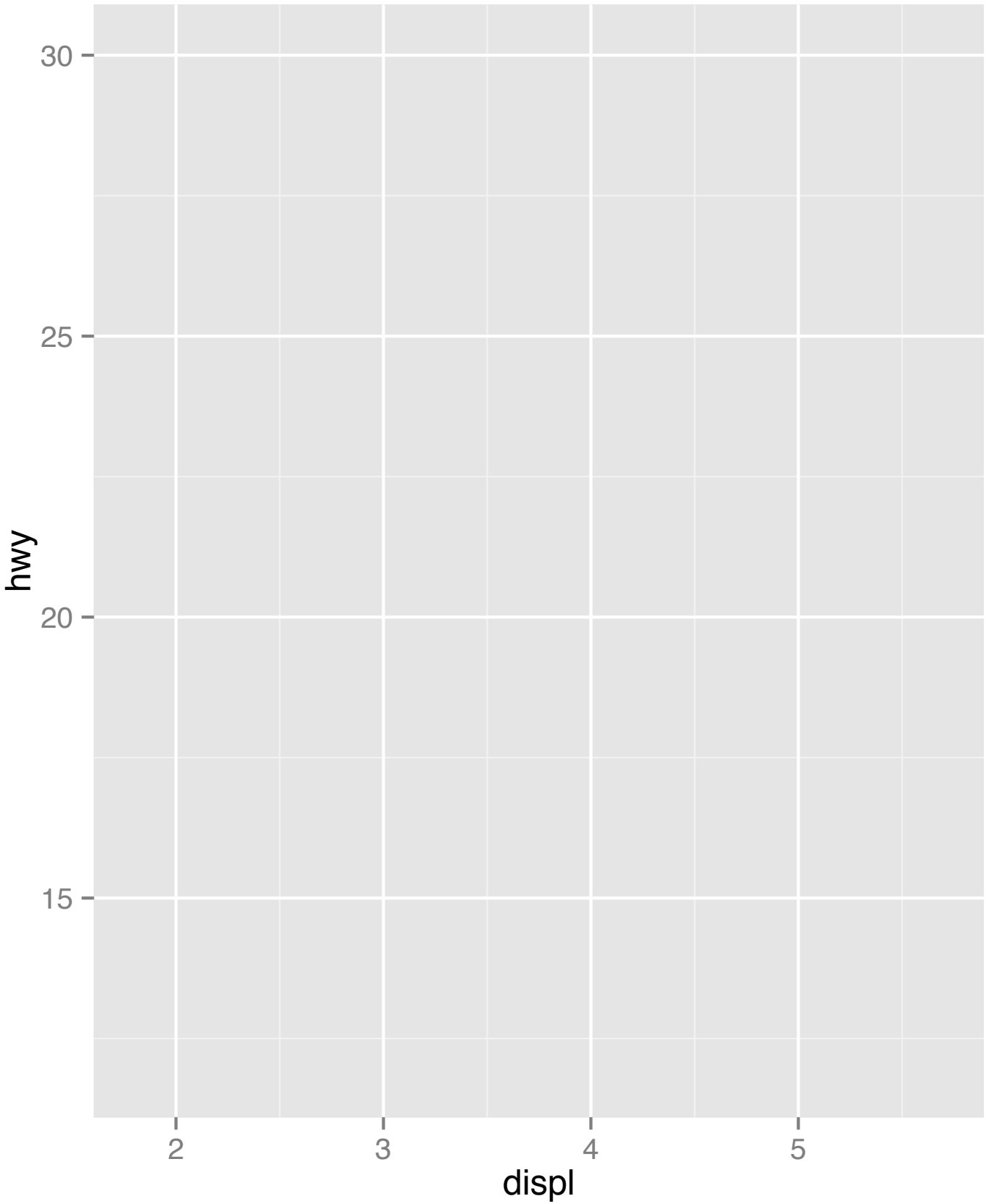
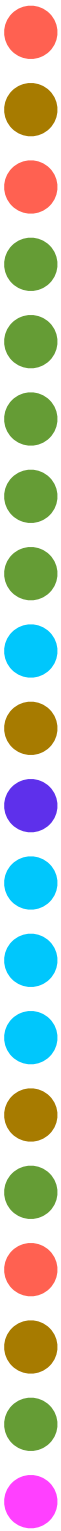
Data

Geom

Coordinate system

# Aesthetic mappings

color			
hwy	displ	cyl	class
17	5.0	8	suv
20	2.7	4	pickup
17	4.0	6	suv
25	2.8	6	compact
27	3.1	6	compact
30	2.0	4	compact
25	2.8	6	compact
23	2.8	6	compact
26	3.0	6	midsize
17	5.4	8	pickup
28	2.5	5	subcompact
29	3.5	6	midsize
26	2.4	4	midsize
29	2.0	4	midsize
15	5.4	8	pickup
29	1.8	4	compact
18	5.7	8	suv
12	4.7	8	pickup
26	2.8	6	compact
24	3.3	6	minivan



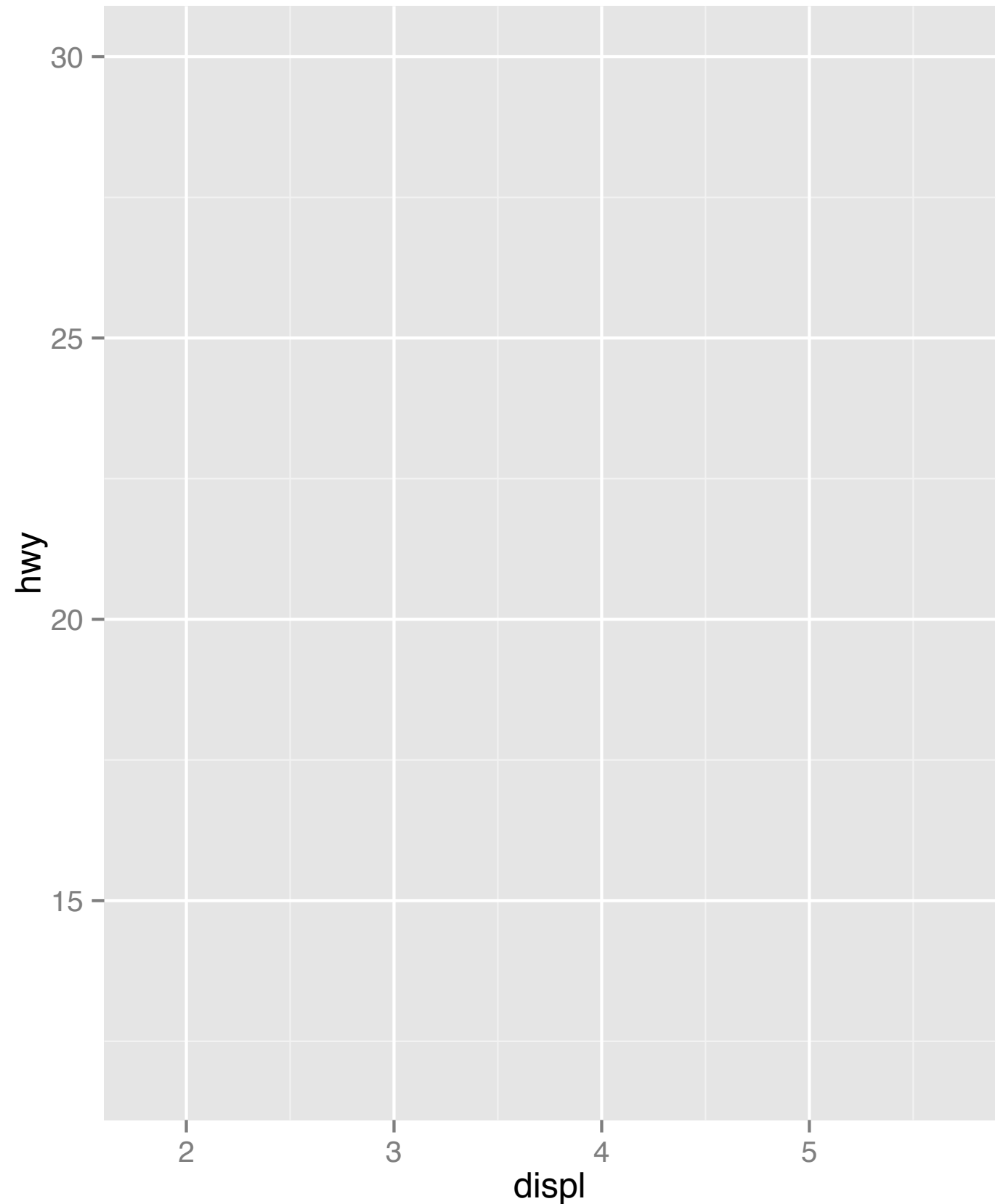
Data

Geom

Coordinate system

# Aesthetic mappings

y	x	color	
hwy	displ	cyl	class
17	5.0	8	suv
20	2.7	4	pickup
17	4.0	6	suv
25	2.8	6	compact
27	3.1	6	compact
30	2.0	4	compact
25	2.8	6	compact
23	2.8	6	compact
26	3.0	6	midsize
17	5.4	8	pickup
28	2.5	5	subcompact
29	3.5	6	midsize
26	2.4	4	midsize
29	2.0	4	midsize
15	5.4	8	pickup
29	1.8	4	compact
18	5.7	8	suv
12	4.7	8	pickup
26	2.8	6	compact
24	3.3	6	minivan



Data

Geom

Coordinate system



# Aesthetic mappings

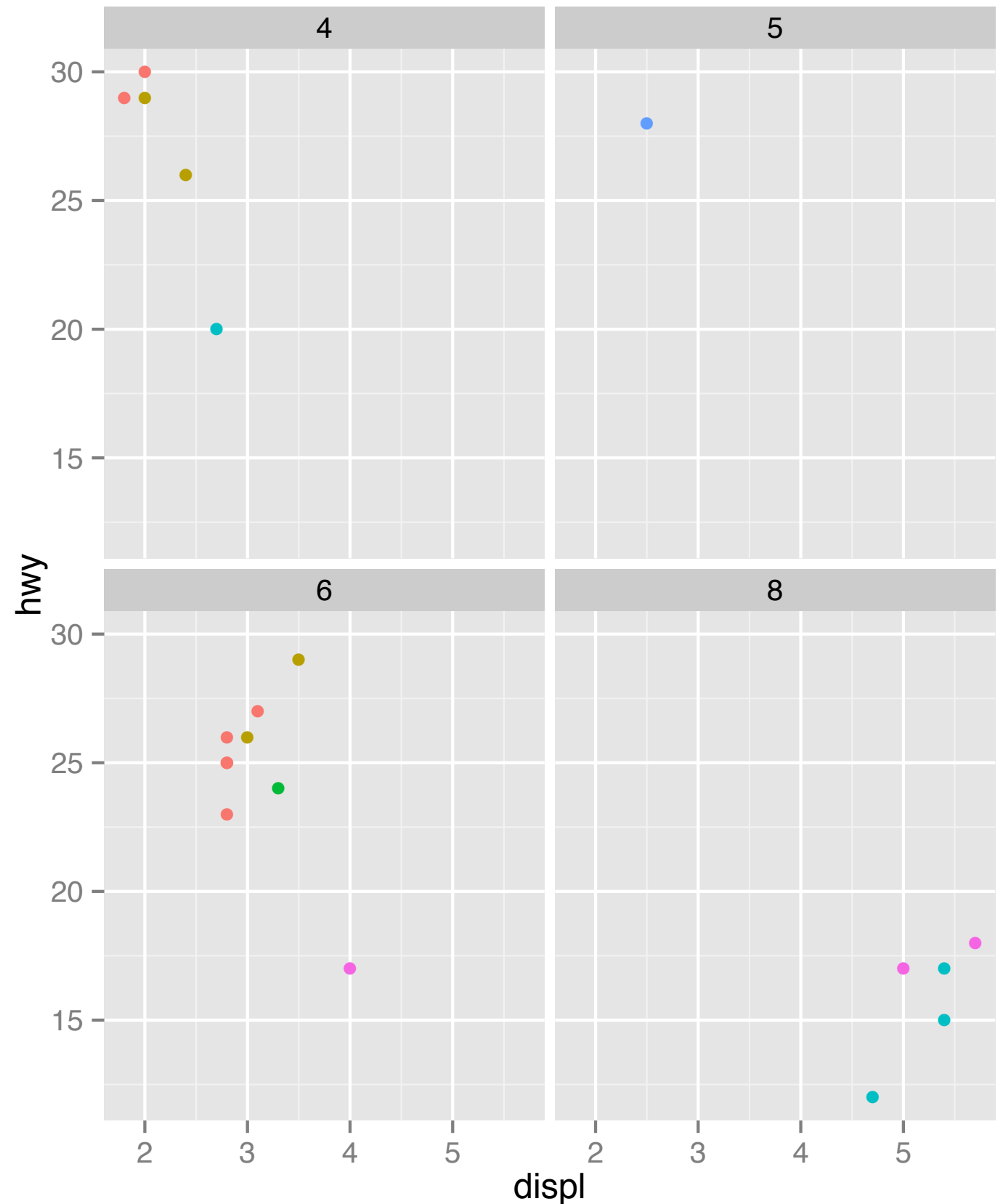
# Facet (or not)

y	x	color	
hwy	displ	cyl	class
17	5.0	8	suv
20	2.7	4	pickup
17	4.0	6	suv
25	2.8	6	compact
27	3.1	6	compact
30	2.0	4	compact
25	2.8	6	compact
23	2.8	6	compact
26	3.0	6	midsize
17	5.4	8	pickup
28	2.5	5	subcompact
29	3.5	6	midsize
26	2.4	4	midsize
29	2.0	4	midsize
15	5.4	8	pickup
29	1.8	4	compact
18	5.7	8	suv
12	4.7	8	pickup
26	2.8	6	compact
24	3.3	6	minivan



Data

Geom



Coordinate system

# What is a plot?

Coordinate system

+ geom

+ data

+ aesthetic mappings

+ position adjustment

This is the grammar  
of graphics\*

# Your turn

Use qplot to create each of the plots described below

Plot 1

**data** = economics

**geom** = line

**aesthetic mappings:** x = date, y = unemploy

Plot 2

**data** = mpg

**geom** = point

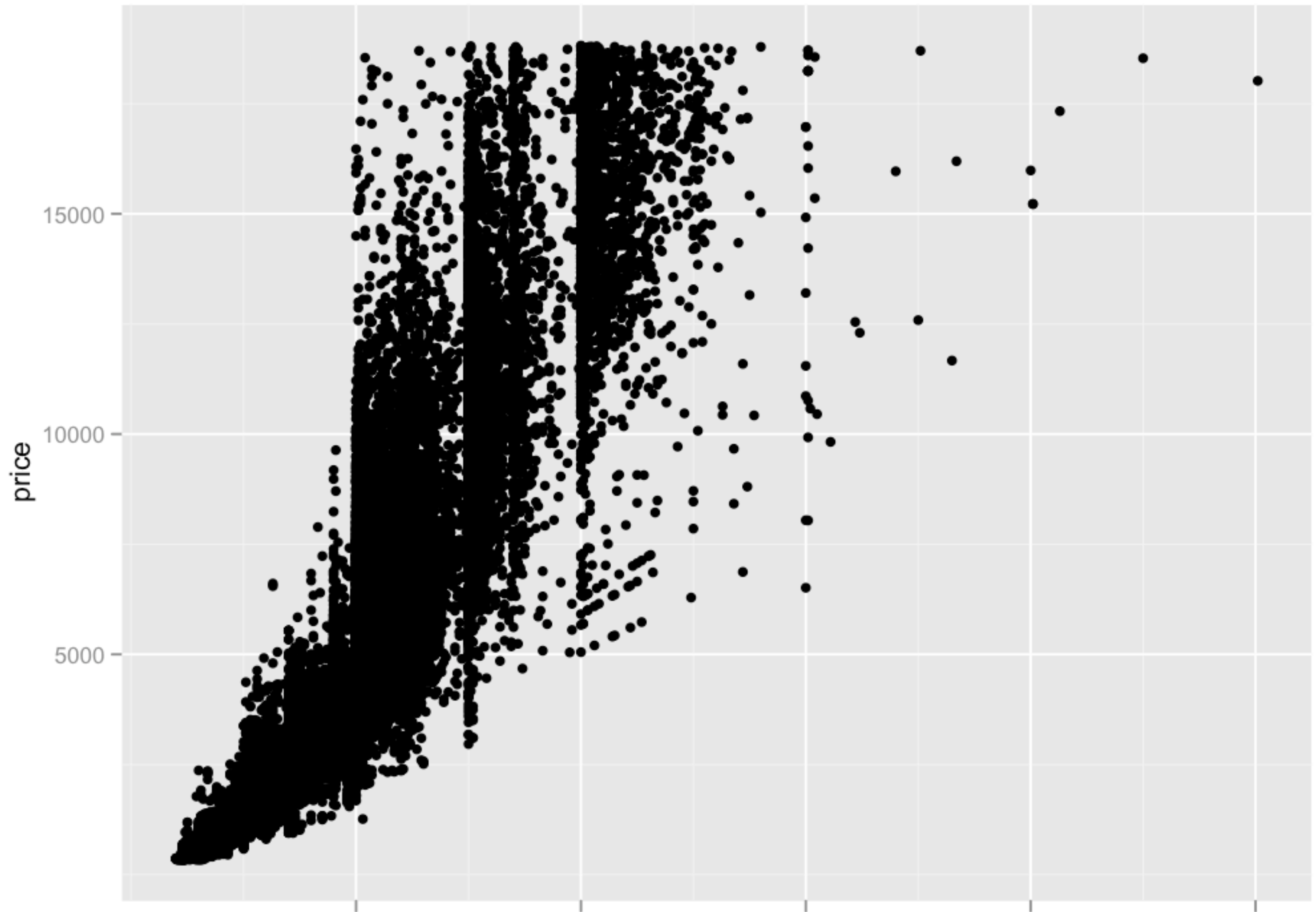
**position** = jitter

**aesthetic mappings:** x = class, y = hwy, color = class

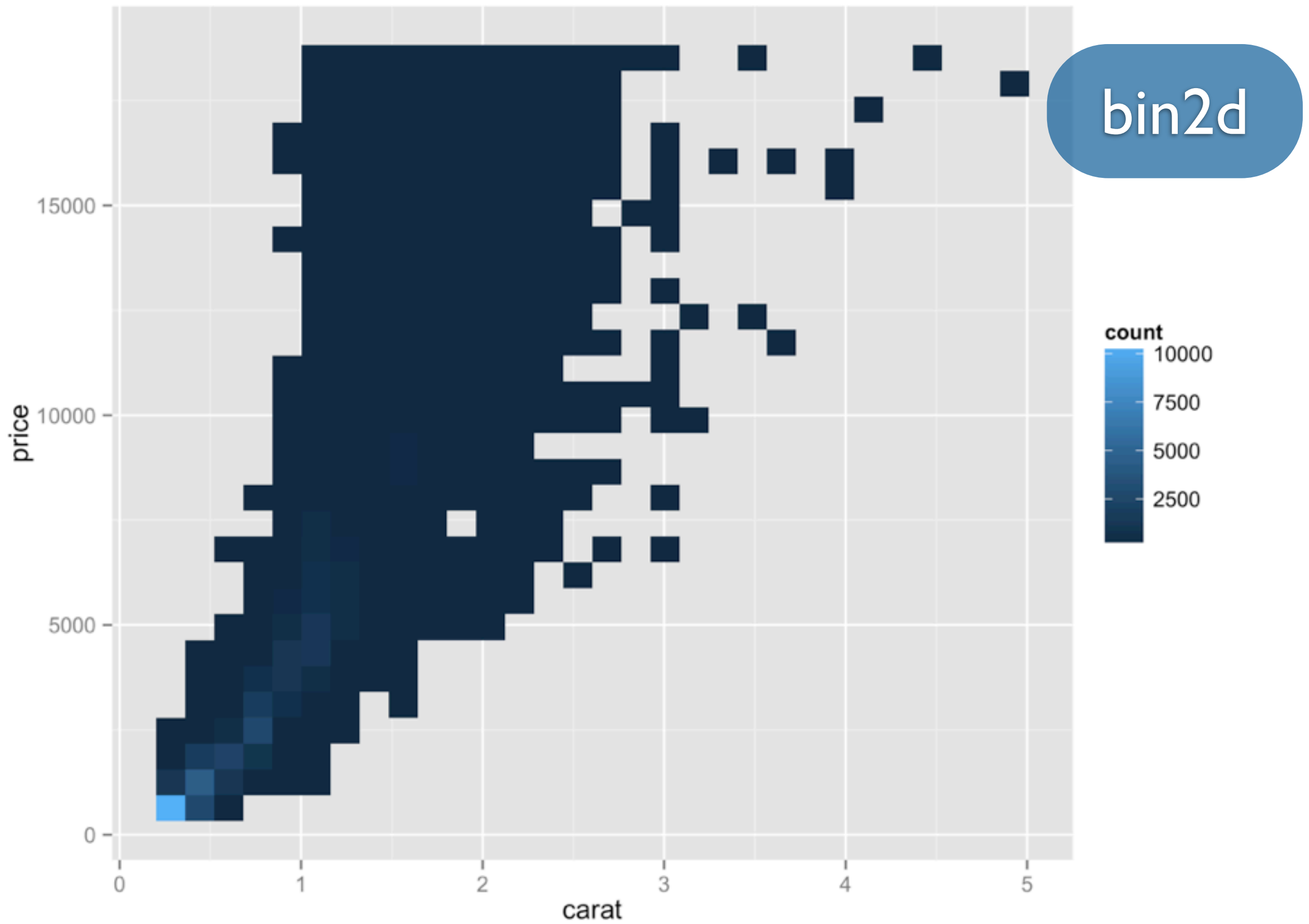
```
qplot(date, unemploy, data = economics,  
      geom = "line")
```

```
qplot(class, hwy, data = mpg,  
      position = "jitter", color = class)
```

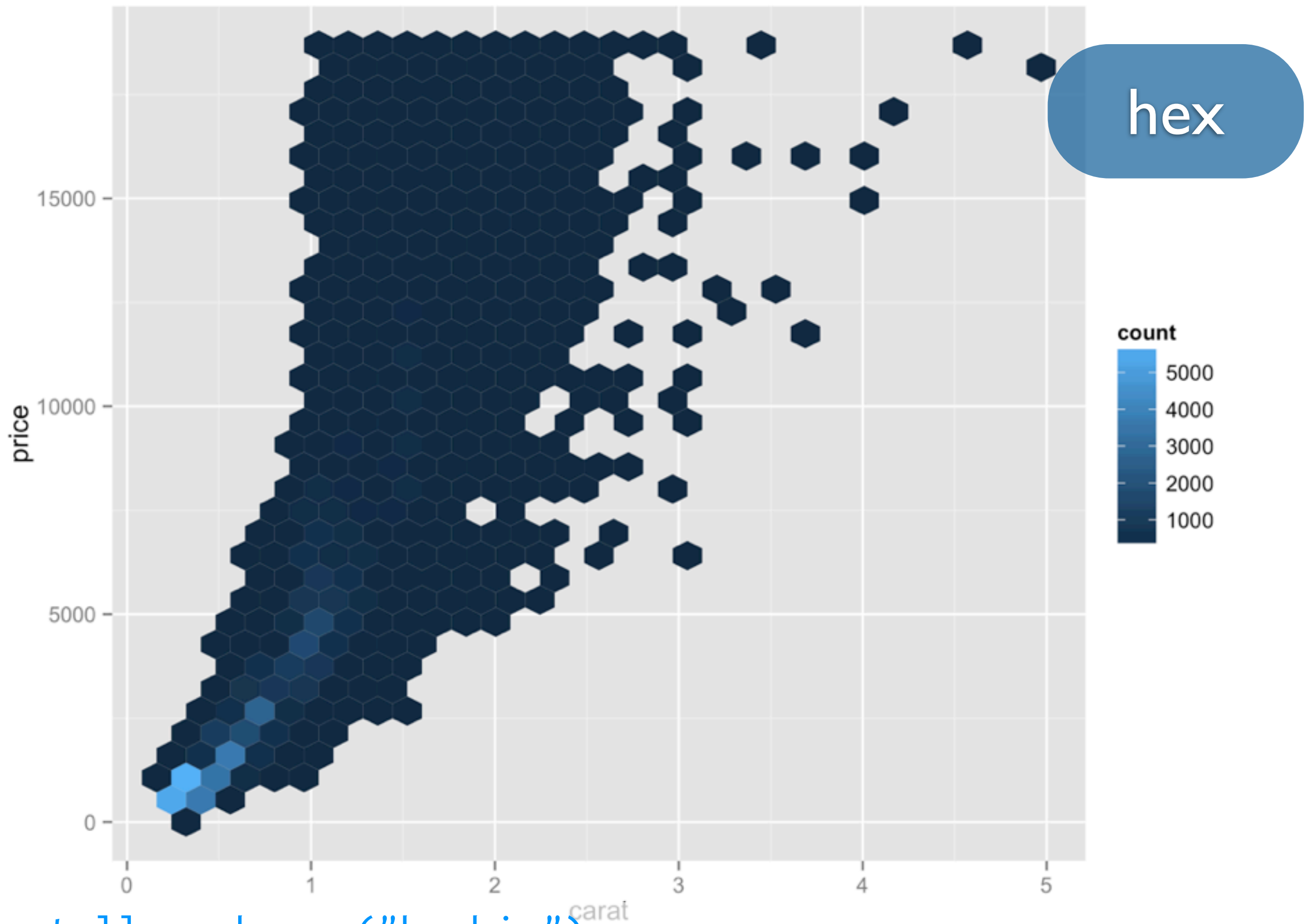
# **Geoms for Big Data**



```
qplot(carat, price, data = diamonds)
```

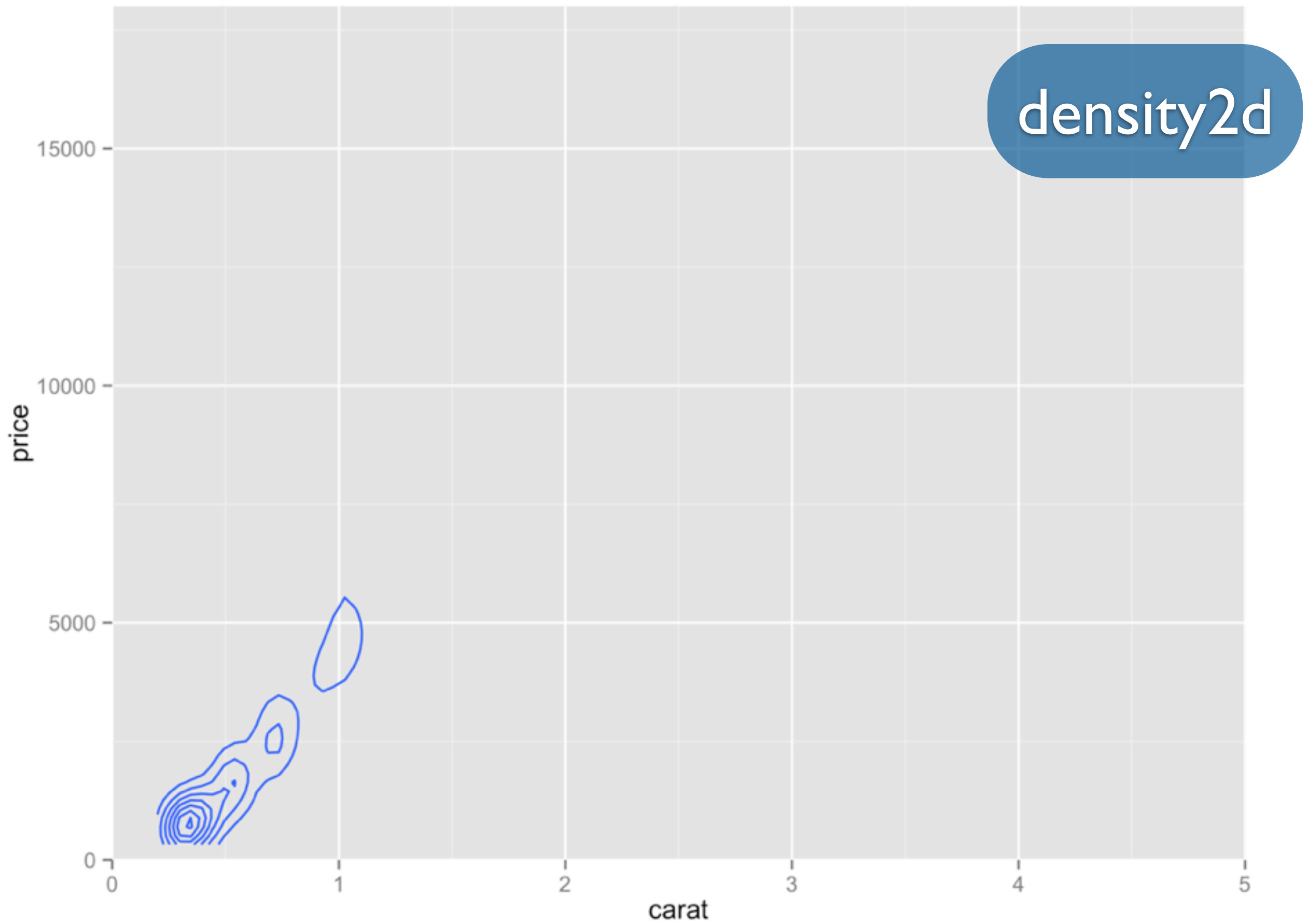


```
qplot(carat, price, data = diamonds, geom = "bin2d")
```

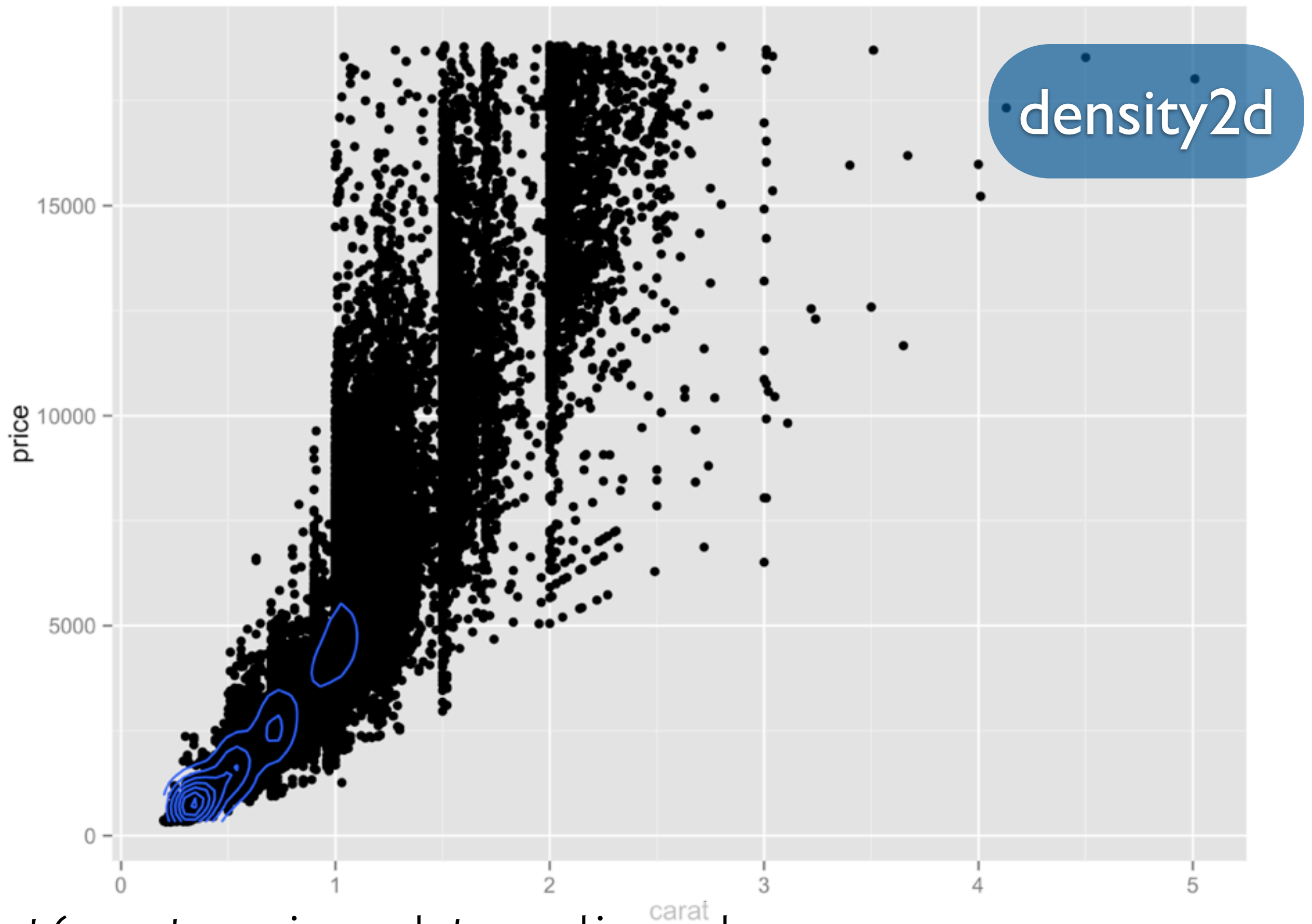


```
# install.packages("hexbin")  
qplot(carat, price, data = diamonds, geom = "hex")
```





```
qplot(carat, price, data = diamonds, geom = "density2d")
```



```
qplot(carat, price, data = diamonds,  
      geom = c("point", "density2d"))
```

## Help topics

### Geoms

Geoms, short for geometric objects, describe the type of plot you will produce.

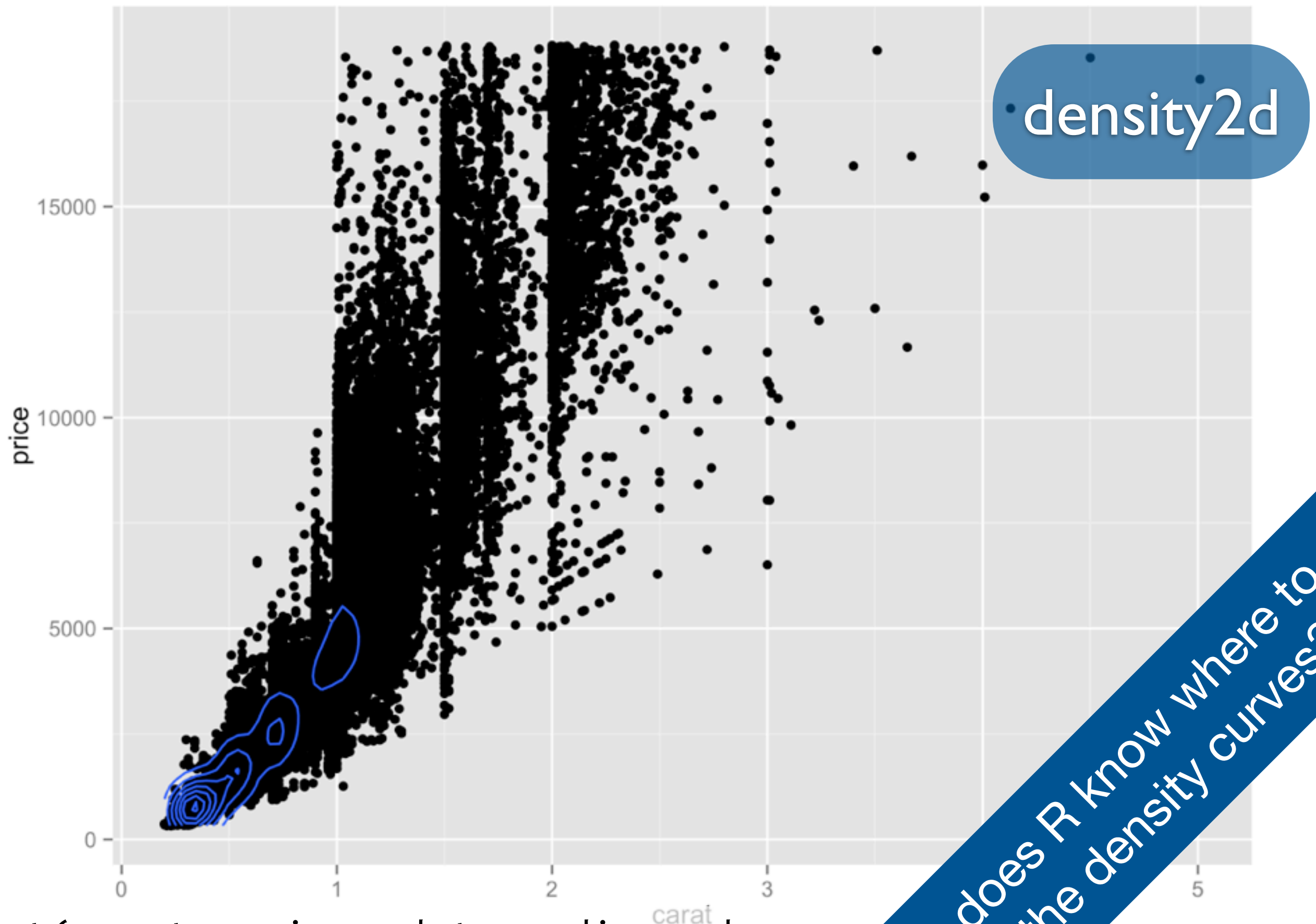
- [geom\\_abline](#)  
Line specified by slope and intercept.
- [geom\\_area](#)  
Area plot.
- [geom\\_bar](#)  
Bars, rectangles with bases on x-axis
- [geom\\_bin2d](#)  
Add heatmap of 2d bin counts.
- [geom\\_blank](#)  
Blank, draws nothing.
- [geom\\_boxplot](#)  
Box and whiskers plot.
- [geom\\_contour](#)  
Display contours of a 3d surface in 2d.
- [geom\\_crossbar](#)  
Hollow bar with middle indicated by horizontal line.
- [geom\\_density](#)  
Display a smooth density estimate.
- [geom\\_density2d](#)  
Contours from a 2d density estimate.
- [geom\\_dotplot](#)  
Dot plot
- [geom\\_errorbar](#)



## Dependencies

- **Depends:** stats, methods
- **Imports:** plyr, digest, grid, gtable, reshape2, scales, memoise, proto, MASS
- **Suggests:** quantreg, Hmisc, mapproj, maps, hexbin, maptools, multcomp, nlme, testthat
- **Extends:** sp

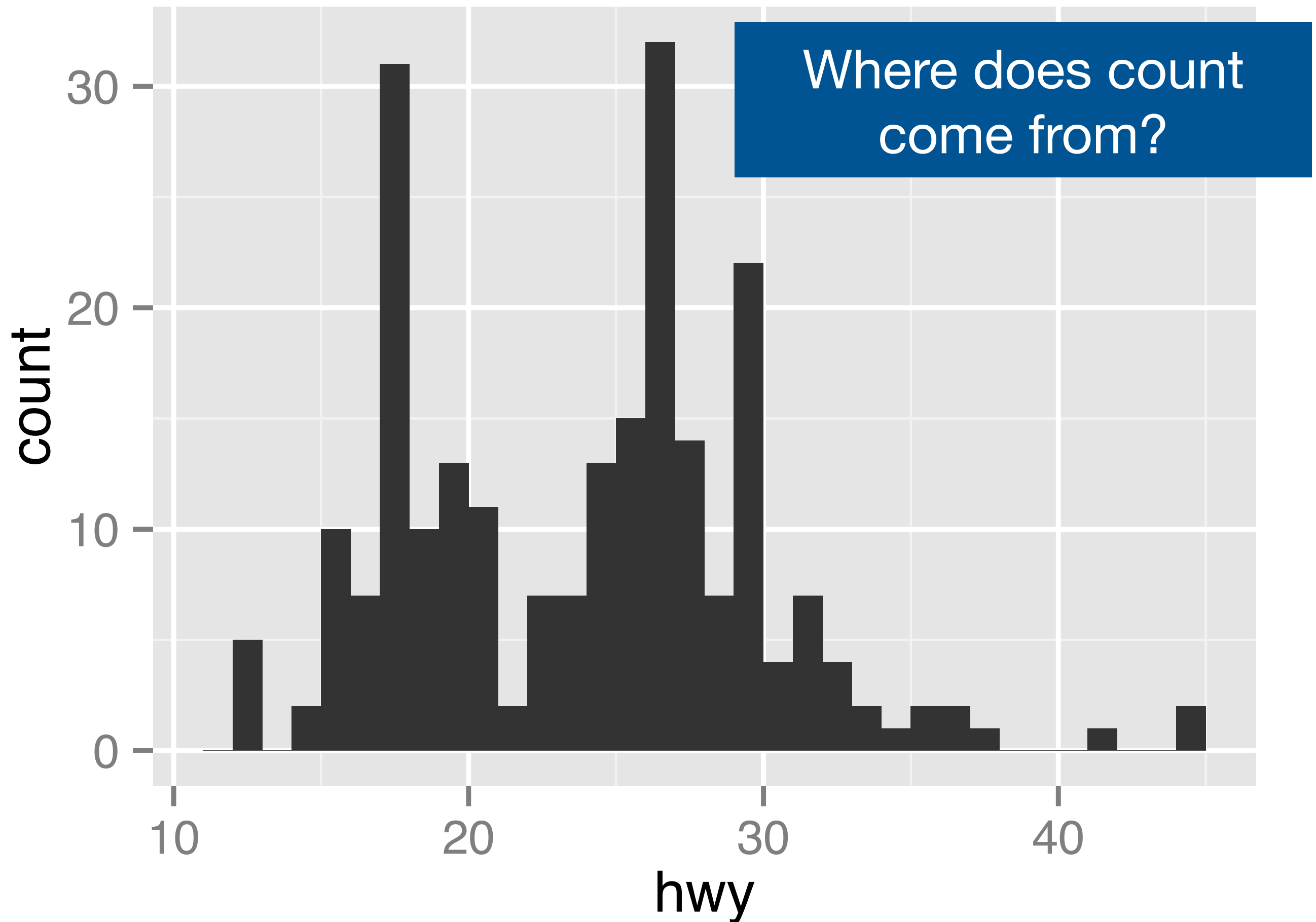
<http://docs.ggplot2.org/current/>



density2d

```
qplot(carat, price, data = diamonds,  
       geom = c("point", "density2d"))
```

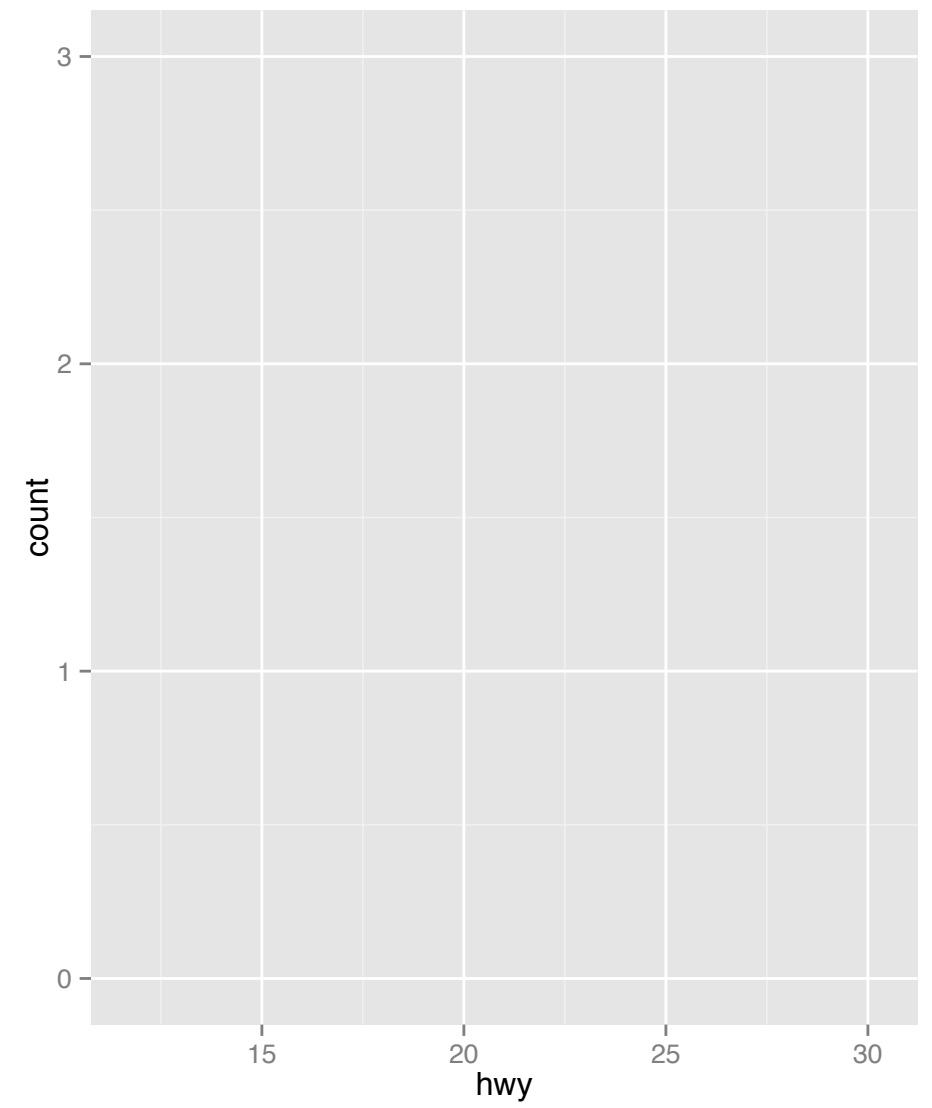
How does R know where to  
draw the density curves?



```
qplot(hwy, data = mpg, geom = "histogram", binwidth = 1)
```

# How to build a plot 2

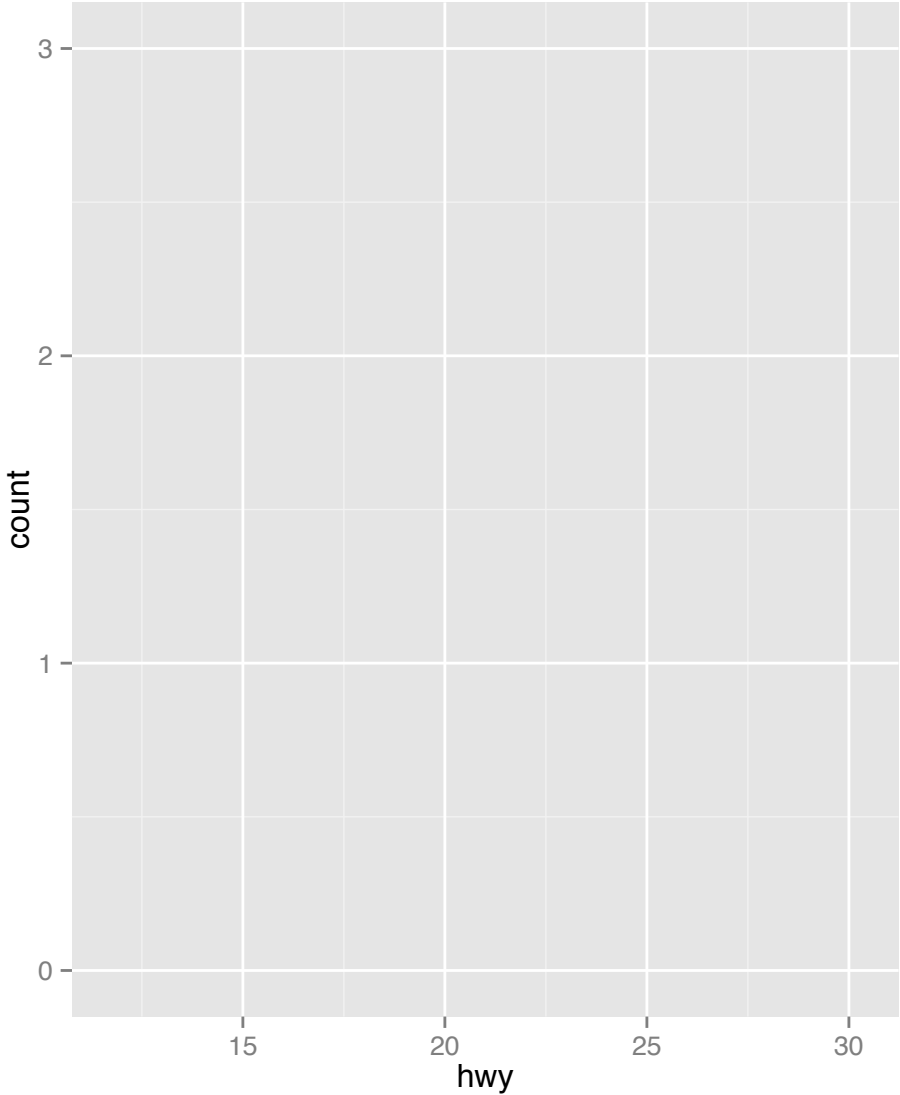
## Stats



Coordinate system

hwy	displ	cyl	class
12	4.7	8	pickup
15	5.4	8	pickup
17	5.0	8	suv
17	4.0	6	suv
17	5.4	8	pickup
18	5.7	8	suv
20	2.7	4	pickup
23	2.8	6	compact
24	3.3	6	minivan
25	2.8	6	compact
25	2.8	6	compact
26	3.0	6	midsize
26	2.4	4	midsize
26	2.8	6	compact
27	3.1	6	compact
28	2.5	5	subcomp
29	3.5	6	midsize
29	2.0	4	midsize
29	1.8	4	compact
30	2.0	4	compact

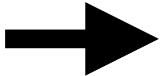
Data



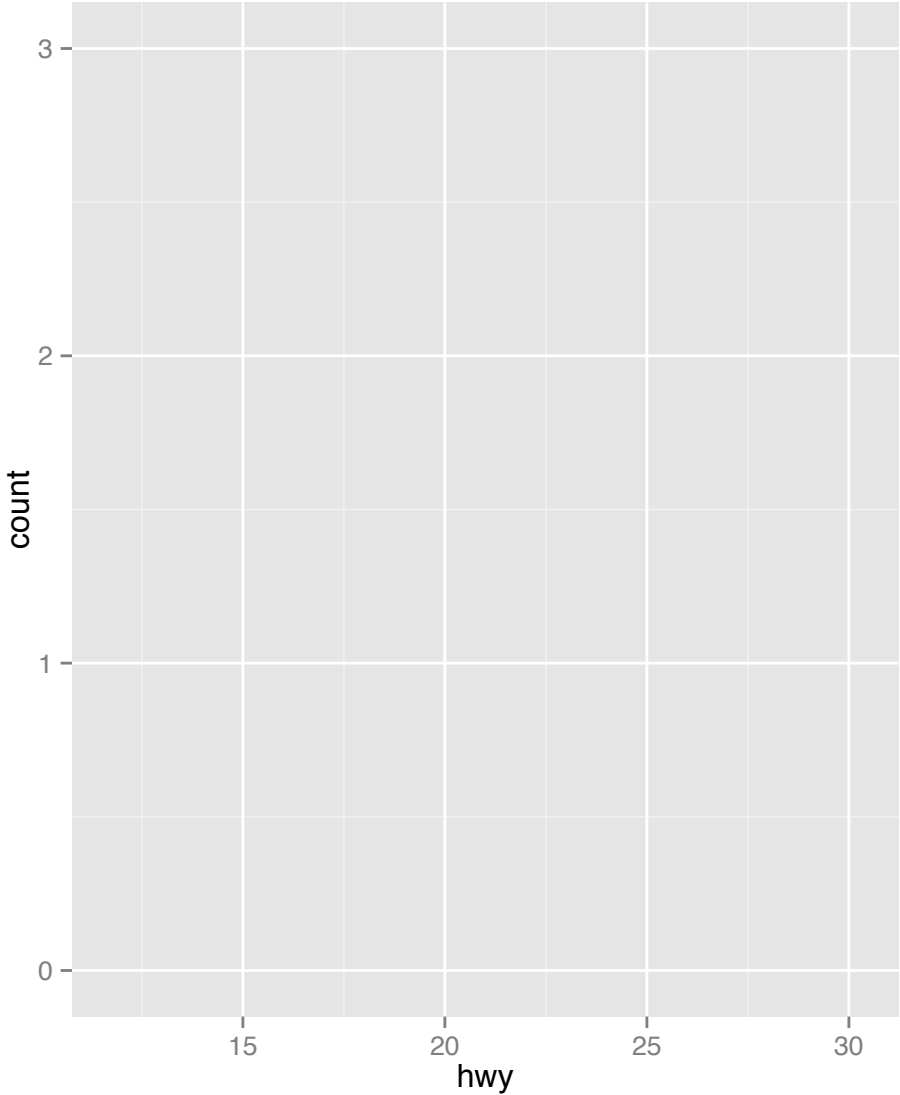
Coordinate system



hwy	displ	cyl	class
12	4.7	8	pickup
15	5.4	8	pickup
17	5.0	8	suv
17	4.0	6	suv
17	5.4	8	pickup
18	5.7	8	suv
20	2.7	4	pickup
23	2.8	6	compact
24	3.3	6	minivan
25	2.8	6	compact
25	2.8	6	compact
26	3.0	6	midsize
26	2.4	4	midsize
26	2.8	6	compact
27	3.1	6	compact
28	2.5	5	subcomp
29	3.5	6	midsize
29	2.0	4	midsize
29	1.8	4	compact
30	2.0	4	compact



bin	hwy	count
1	12	1
2	15	1
3	17	3
4	18	1
5	20	1
6	23	1
7	24	1
8	25	2
9	26	3
10	27	1
11	28	1
12	29	3
13	30	1

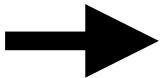


Data

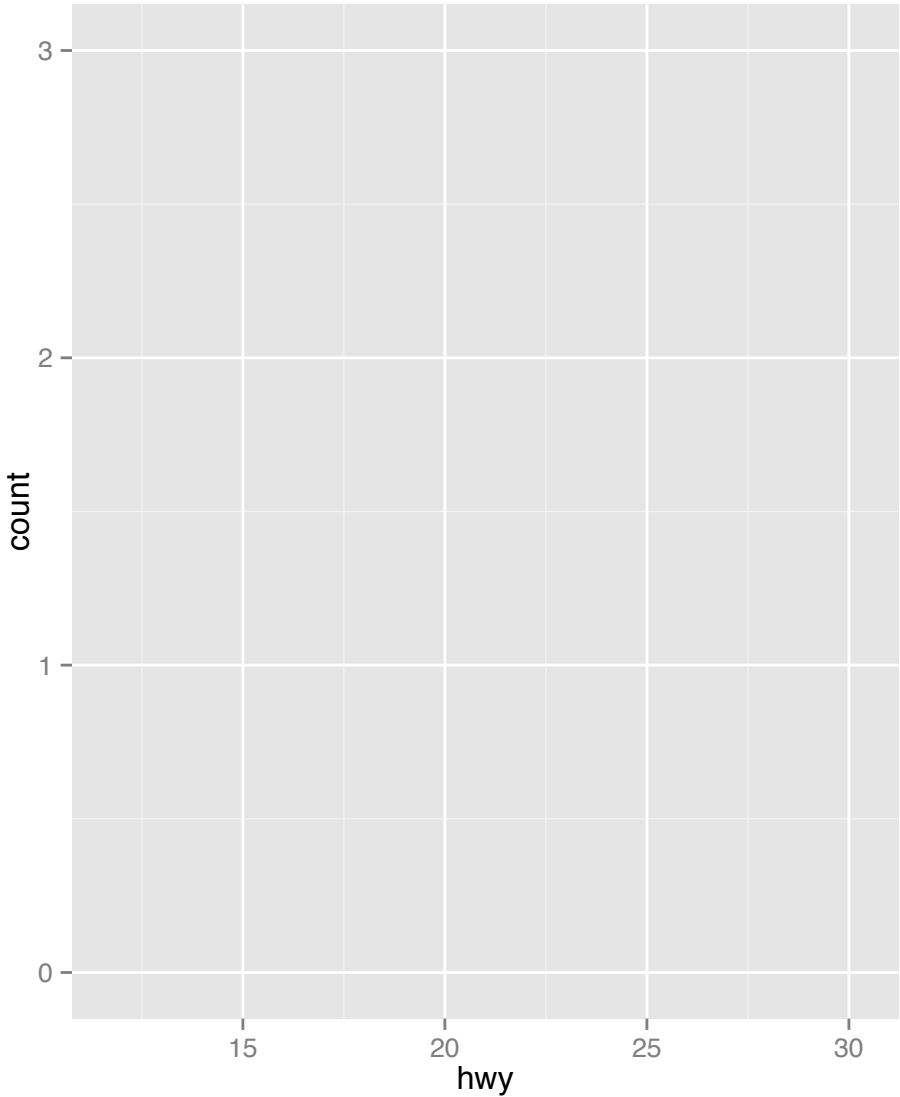
Stat

Coordinate system

hwy	displ	cyl	class
12	4.7	8	pickup
15	5.4	8	pickup
17	5.0	8	suv
17	4.0	6	suv
17	5.4	8	pickup
18	5.7	8	suv
20	2.7	4	pickup
23	2.8	6	compact
24	3.3	6	minivan
25	2.8	6	compact
25	2.8	6	compact
26	3.0	6	midsize
26	2.4	4	midsize
26	2.8	6	compact
27	3.1	6	compact
28	2.5	5	subcomp
29	3.5	6	midsize
29	2.0	4	midsize
29	1.8	4	compact
30	2.0	4	compact



bin	hwy	count
1	12	1
2	15	1
3	17	3
4	18	1
5	20	1
6	23	1
7	24	1
8	25	2
9	26	3
10	27	1
11	28	1
12	29	3
13	30	1



Data

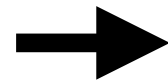
Stat

Geom

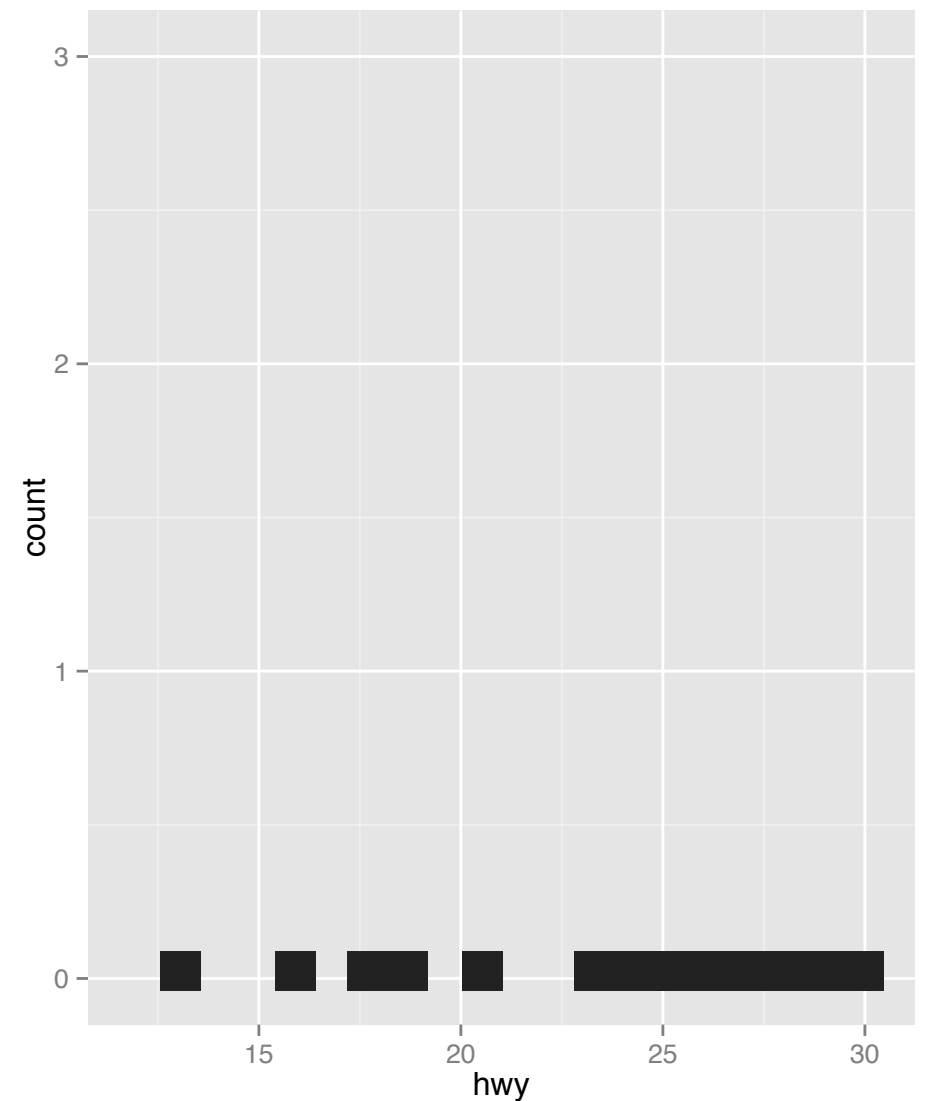
Coordinate system

# Aesthetic mappings

hwy	displ	cyl	class
12	4.7	8	pickup
15	5.4	8	pickup
17	5.0	8	suv
17	4.0	6	suv
17	5.4	8	pickup
18	5.7	8	suv
20	2.7	4	pickup
23	2.8	6	compact
24	3.3	6	minivan
25	2.8	6	compact
25	2.8	6	compact
26	3.0	6	midsize
26	2.4	4	midsize
26	2.8	6	compact
27	3.1	6	compact
28	2.5	5	subcomp
29	3.5	6	midsize
29	2.0	4	midsize
29	1.8	4	compact
30	2.0	4	compact



x			
bin	hwy	count	
1	12	1	■
2	15	1	■
3	17	3	■
4	18	1	■
5	20	1	■
6	23	1	■
7	24	1	■
8	25	2	■
9	26	3	■
10	27	1	■
11	28	1	■
12	29	3	■
13	30	1	■



Data

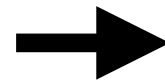
Stat

Geom

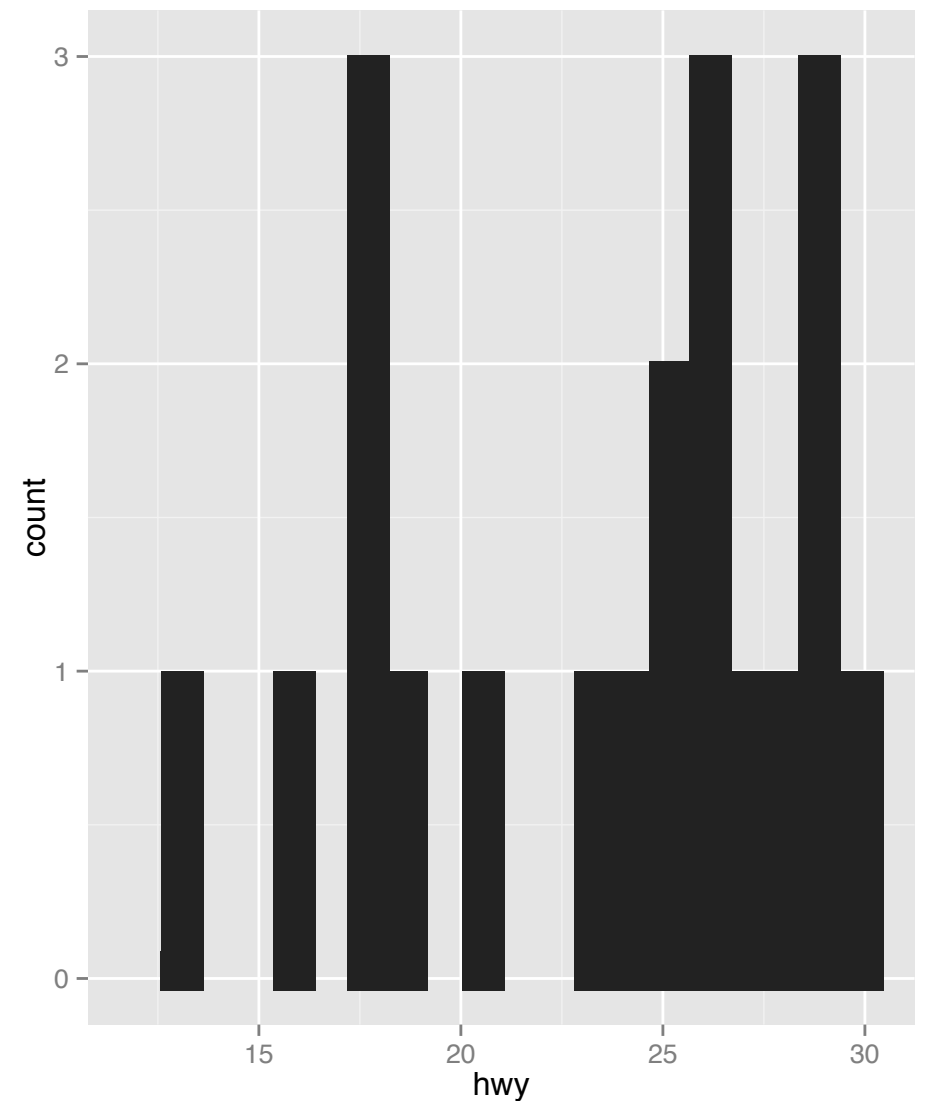
Coordinate system

# Aesthetic mappings

hwy	displ	cyl	class
12	4.7	8	pickup
15	5.4	8	pickup
17	5.0	8	suv
17	4.0	6	suv
17	5.4	8	pickup
18	5.7	8	suv
20	2.7	4	pickup
23	2.8	6	compact
24	3.3	6	minivan
25	2.8	6	compact
25	2.8	6	compact
26	3.0	6	midsize
26	2.4	4	midsize
26	2.8	6	compact
27	3.1	6	compact
28	2.5	5	subcomp
29	3.5	6	midsize
29	2.0	4	midsize
29	1.8	4	compact
30	2.0	4	compact



		x	y	ymax
bin	hwy	count		
1	12	1		■
2	15	1		■
3	17	3		■
4	18	1		■
5	20	1		■
6	23	1		■
7	24	1		■
8	25	2		■
9	26	3		■
10	27	1		■
11	28	1		■
12	29	3		■
13	30	1		■



Data

Stat

Geom

Coordinate system

# Stat

A transformation done to the data before plotting it

# What is a plot?

Coordinate system

+ geom

+ data

+ aesthetic mappings

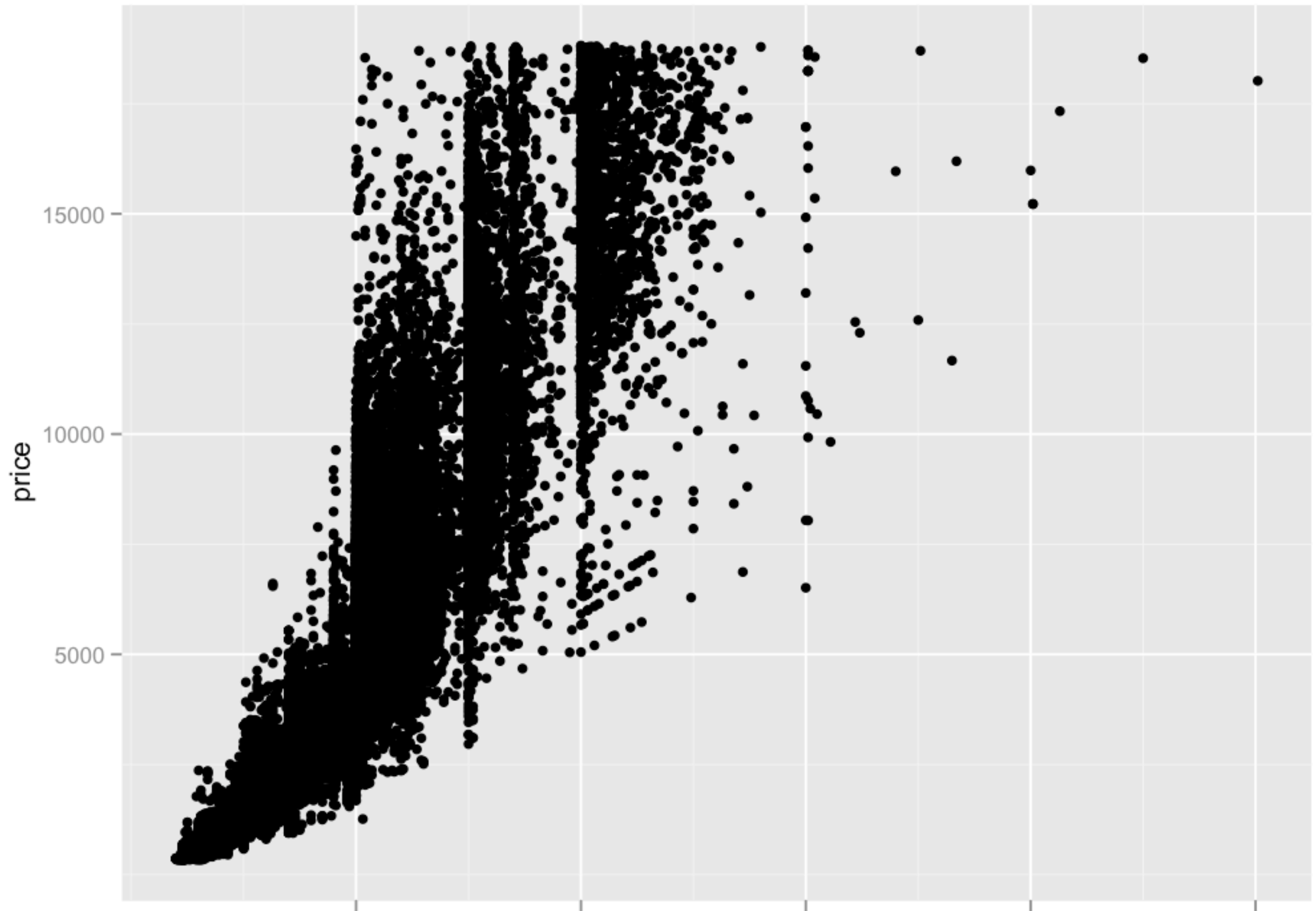
+ position adjustment

+ stat

This is the grammar  
of graphics

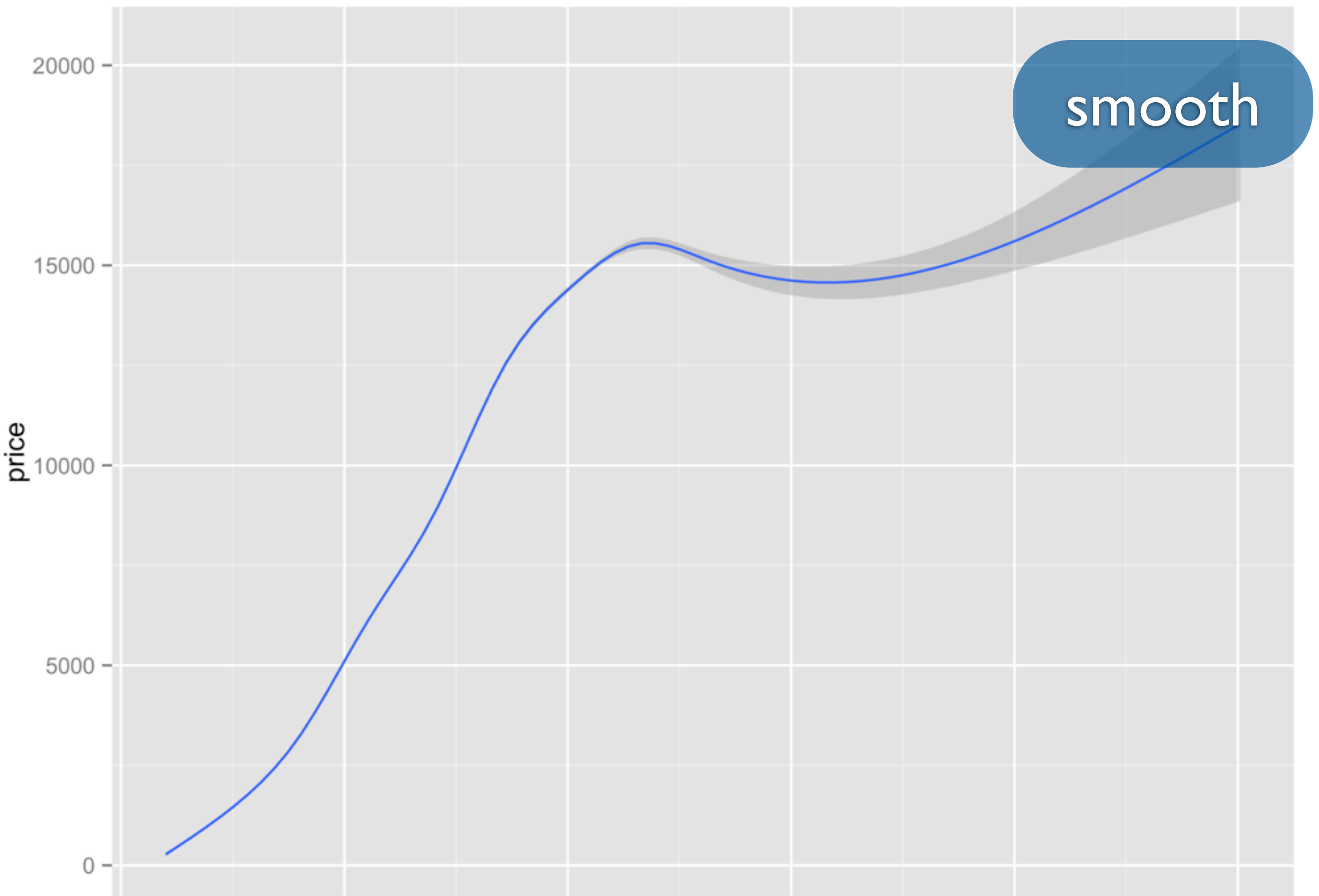
geom	base geom	stat
histogram	bar	bin
smooth	line	smooth
boxplot	boxplot	boxplot
density	line	density
freqpoly	line	

Geoms in ggplot2 know when  
and how to use a stat

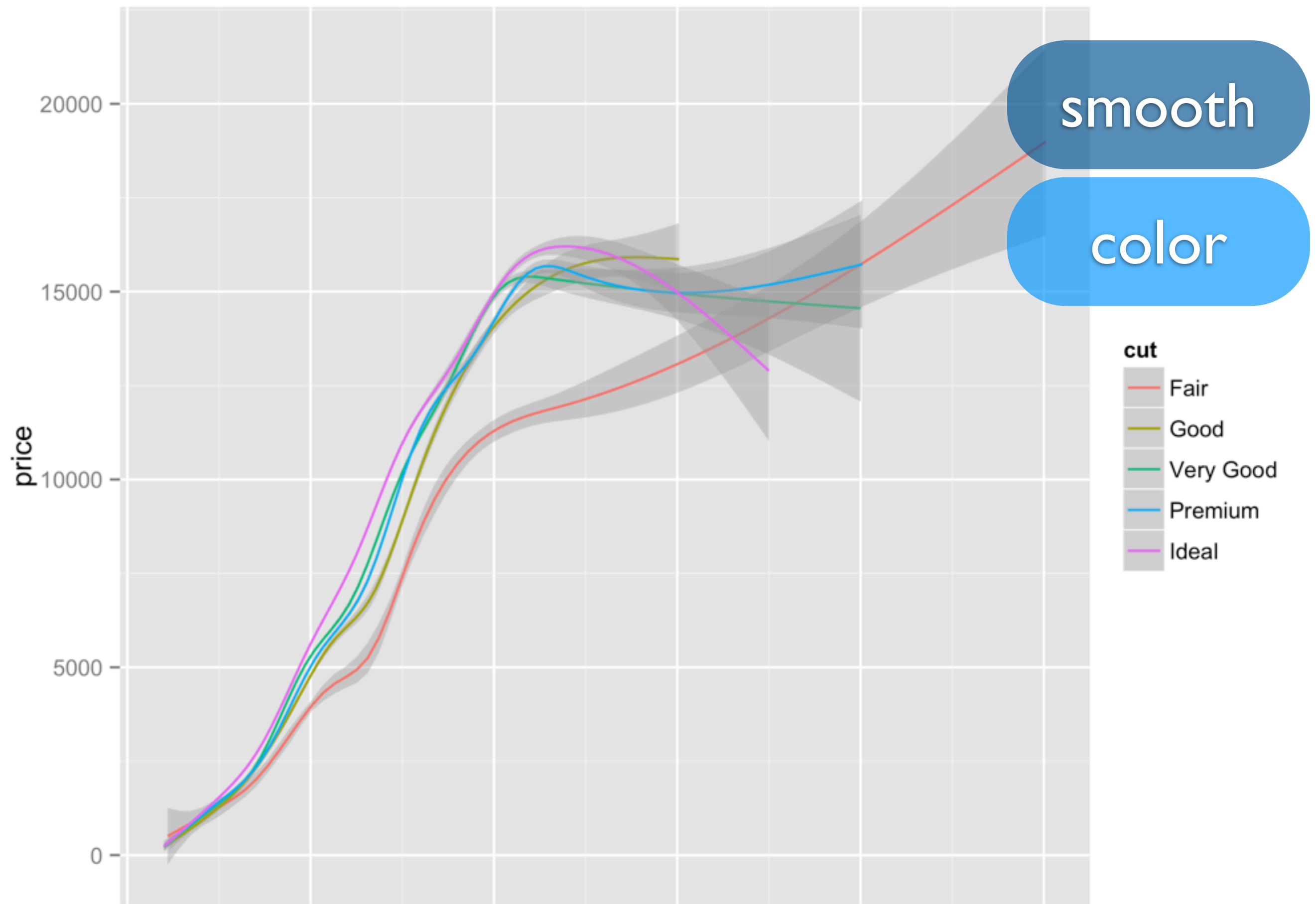


```
qplot(carat, price, data = diamonds)
```

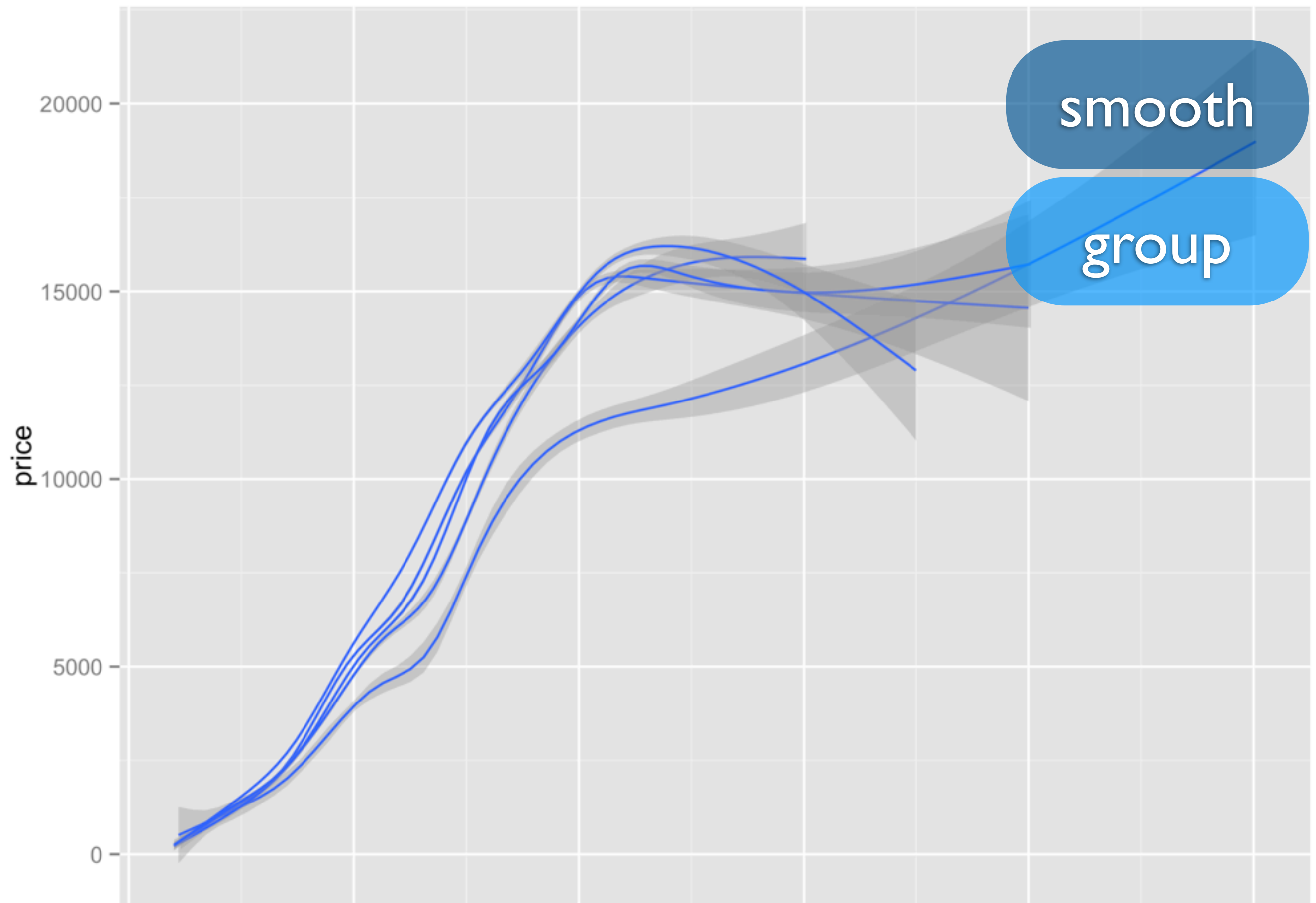




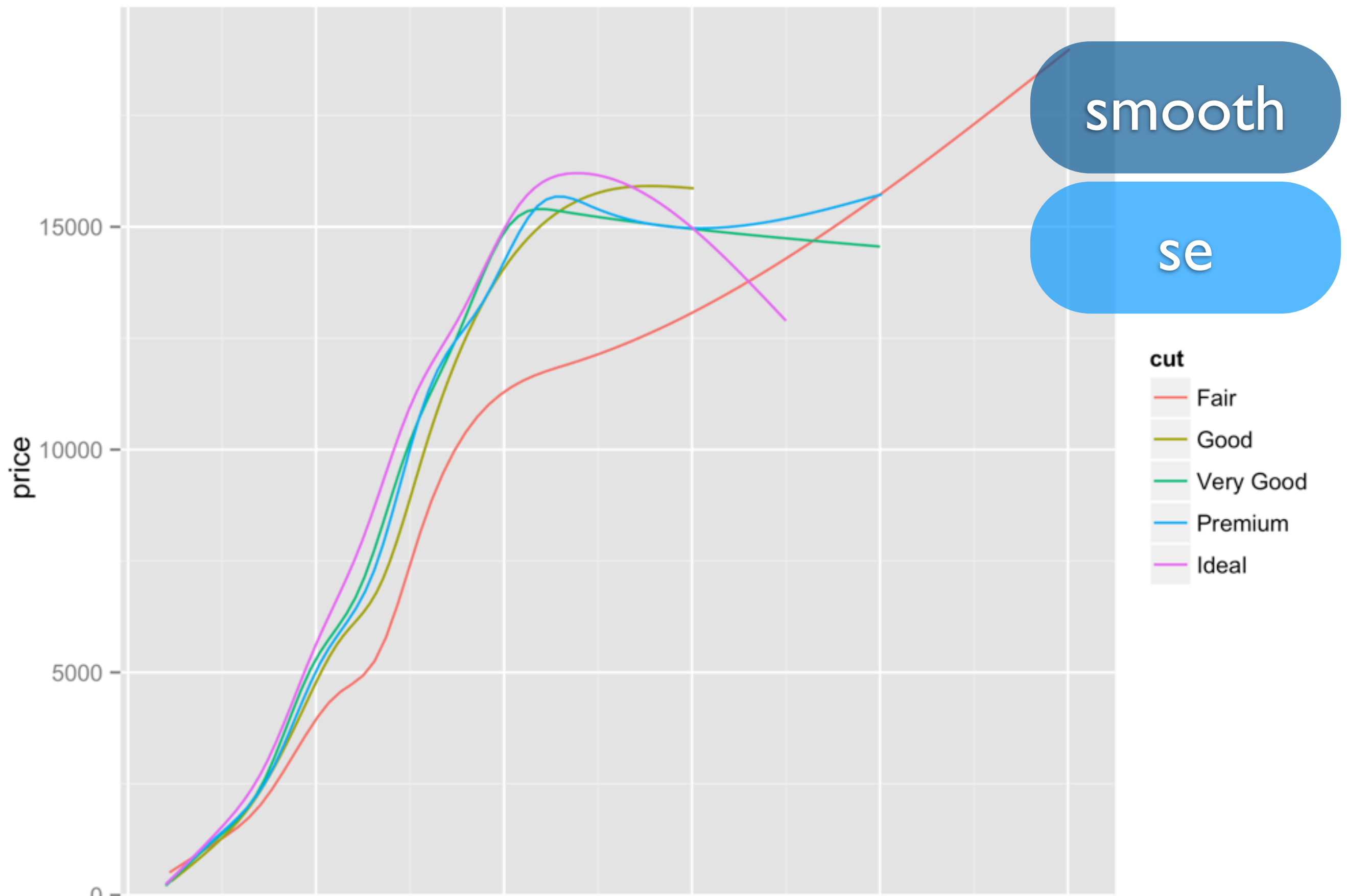
```
qplot(carat, price, data = diamonds, geom = "smooth")
```



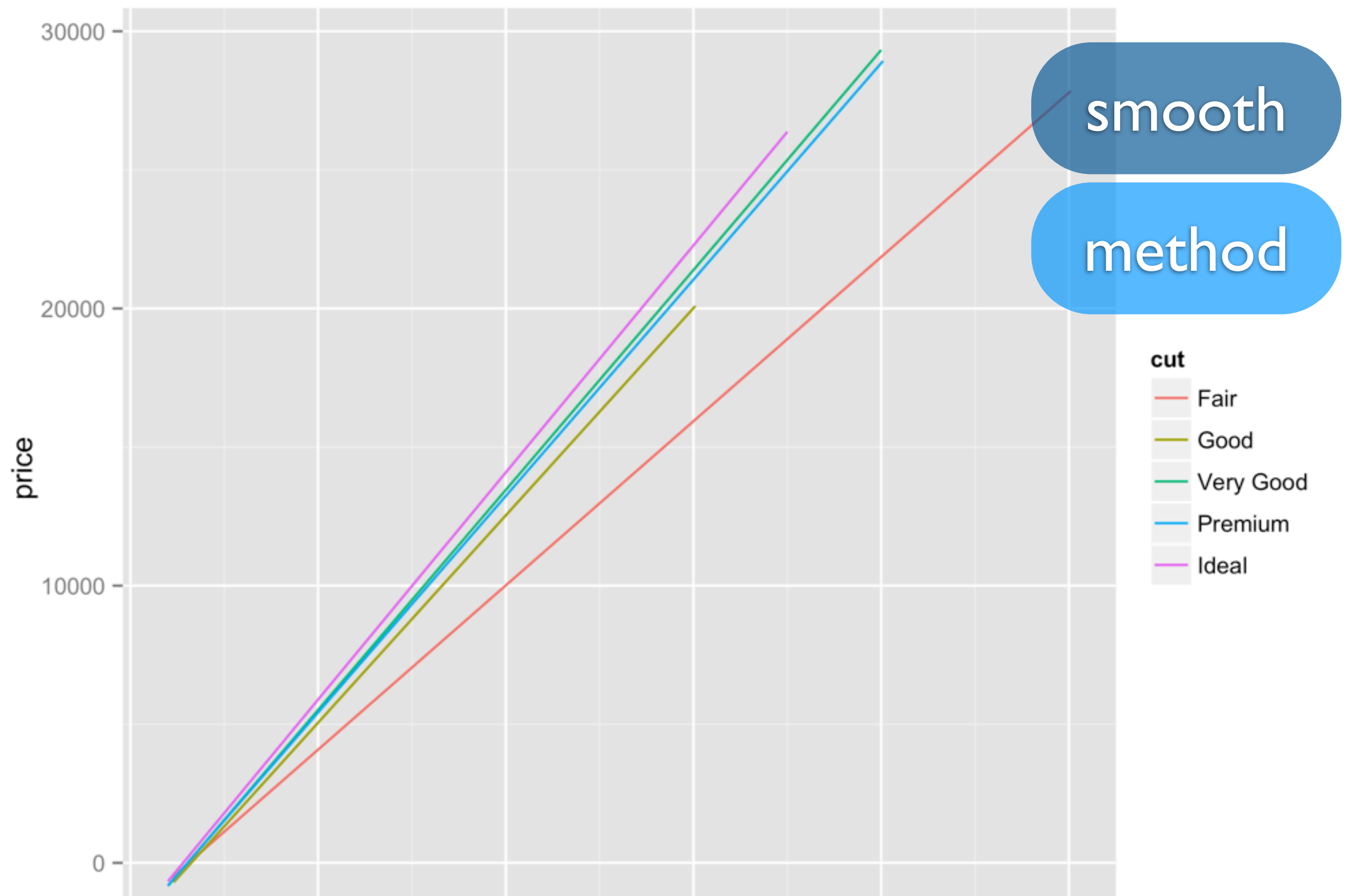
```
qplot(carat, price, data = diamonds, geom = "smooth", color = cut)
```



```
qplot(carat, price, data = diamonds, geom = "smooth", group = cut)
```



```
qplot(carat, price, data = diamonds, geom = "smooth",  
      color = cut, se = FALSE)
```

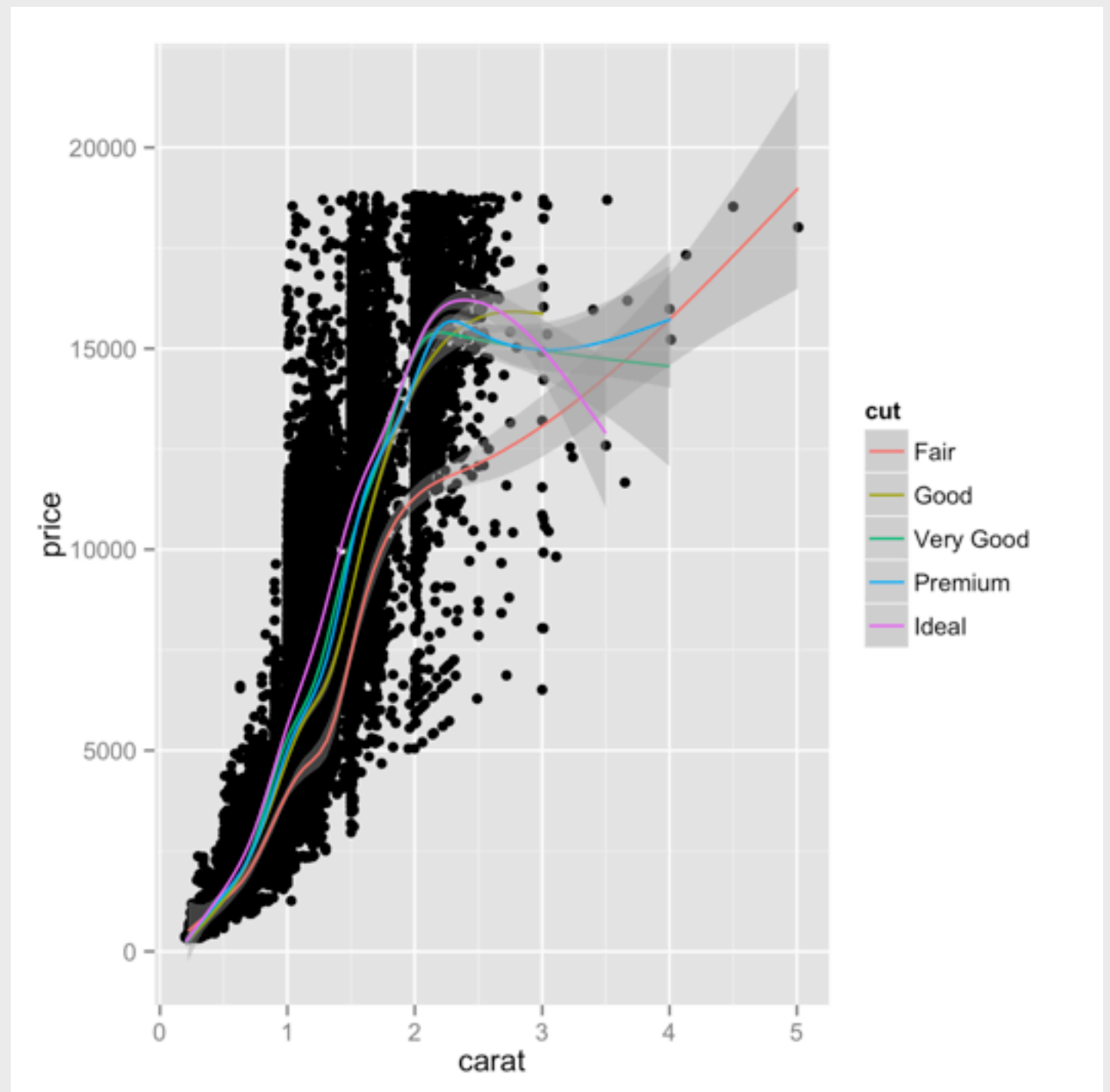


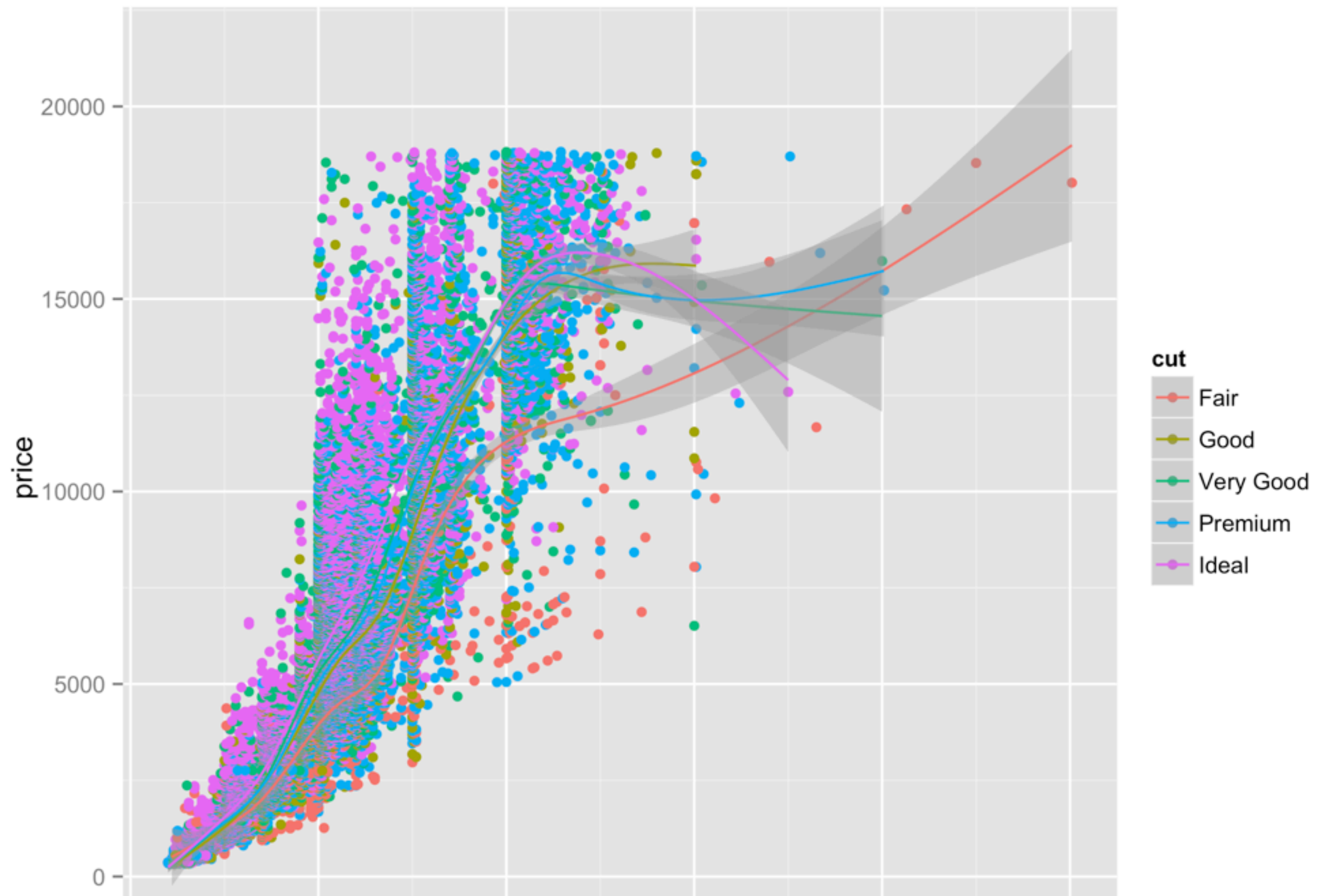
```
qplot(carat, price, data = diamonds, geom = "smooth",  
       color = cut, se = FALSE, method = lm)
```

# Your turn

It's useful to overlay a summary on top of the raw data.

Can you make this plot?  
If not, why not?





```
qplot(carat, price, data = diamonds, color = cut,  
      geom = c("point", "smooth"))
```

`qplot` is good for making quick plots, but it fails if you want to assign different aesthetics to different geoms.

The solution? layers



# Layers

# Geom functions

There are two ways to add additional geoms to a plot

1) A vector of geom names:

```
qplot(carat, price, data = diamonds,  
      geom = c("point", "smooth"))
```

2) The geom functions

```
qplot(carat, price, data = diamonds) +  
  geom_smooth()
```

Always begins  
with geom\_

geom's name

open and closed  
parentheses

geom\_smooth()

# Advantage 1

Geom functions provide a way to look up help pages for specific geoms

`?geom_smooth`

Or even better

<http://docs.ggplot2.org/current/>

## Help topics

### Geoms

Geoms, short for geometric objects, describe the type of plot you will produce.

- [geom\\_abline](#)  
Line specified by slope and intercept.
- [geom\\_area](#)  
Area plot.
- [geom\\_bar](#)  
Bars, rectangles with bases on x-axis
- [geom\\_bin2d](#)  
Add heatmap of 2d bin counts.
- [geom\\_blank](#)  
Blank, draws nothing.
- [geom\\_boxplot](#)  
Box and whiskers plot.
- [geom\\_contour](#)  
Display contours of a 3d surface in 2d.
- [geom\\_crossbar](#)  
Hollow bar with middle indicated by horizontal line.
- [geom\\_density](#)  
Display a smooth density estimate.
- [geom\\_density2d](#)  
Contours from a 2d density estimate.
- [geom\\_dotplot](#)  
Dot plot



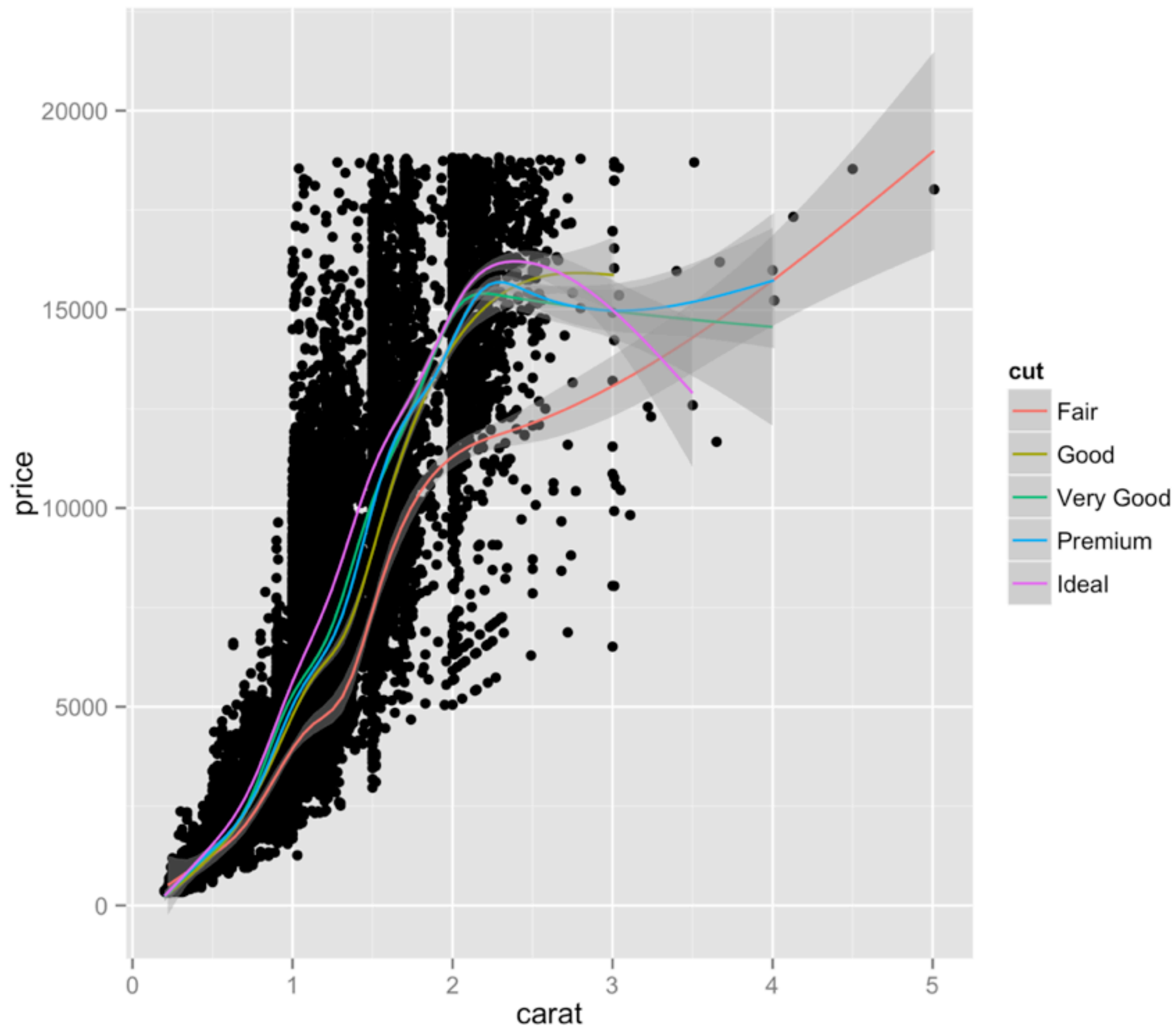
## Dependencies

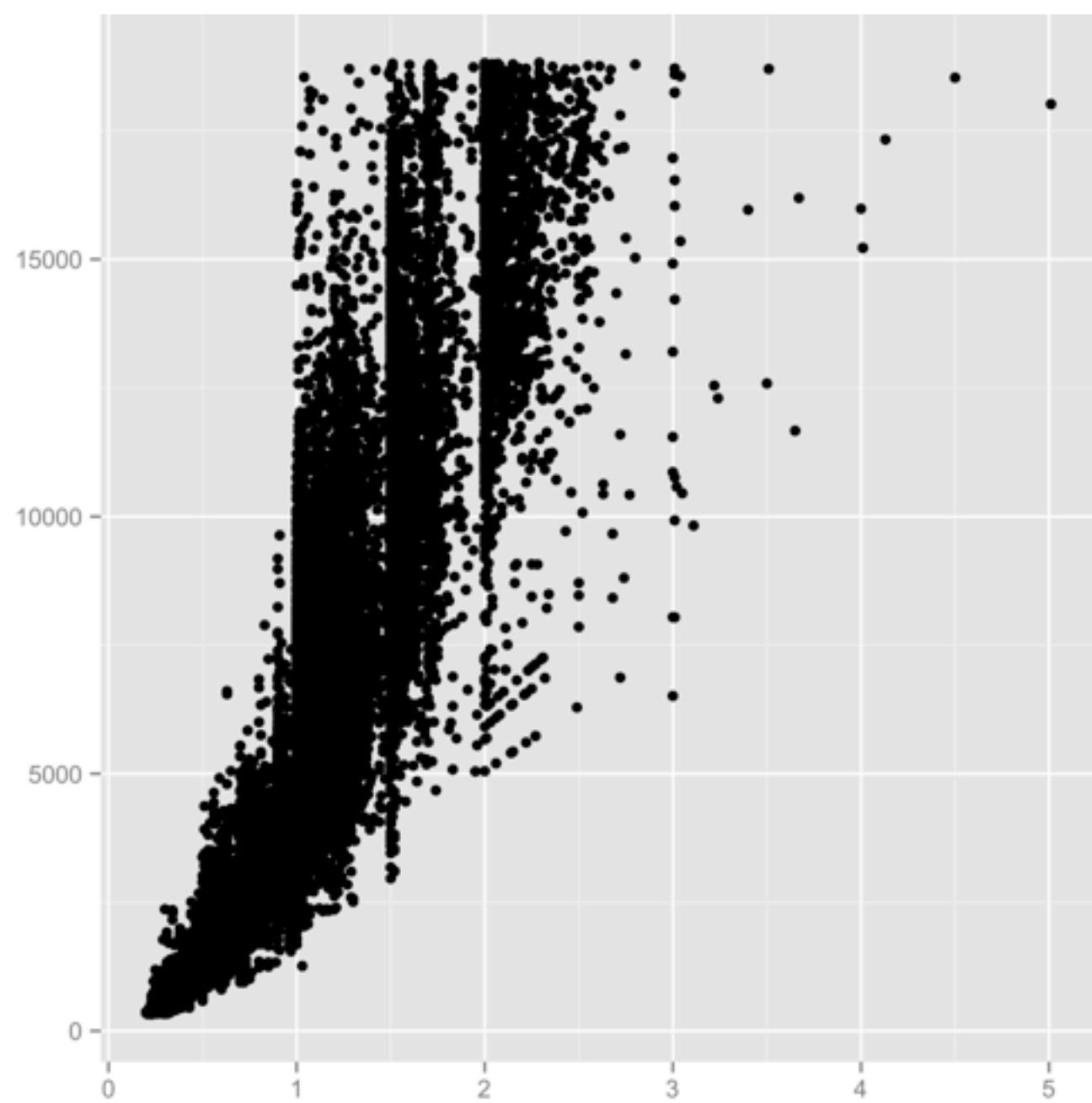
- **Depends:** stats, methods
- **Imports:** plyr, digest, grid, gtable, reshape2, scales, memoise, proto, MASS
- **Suggests:** quantreg, Hmisc, mapproj, maps, hexbin, maptools, multcomp, nlme, testthat
- **Extends:** sp

# Advantage 2

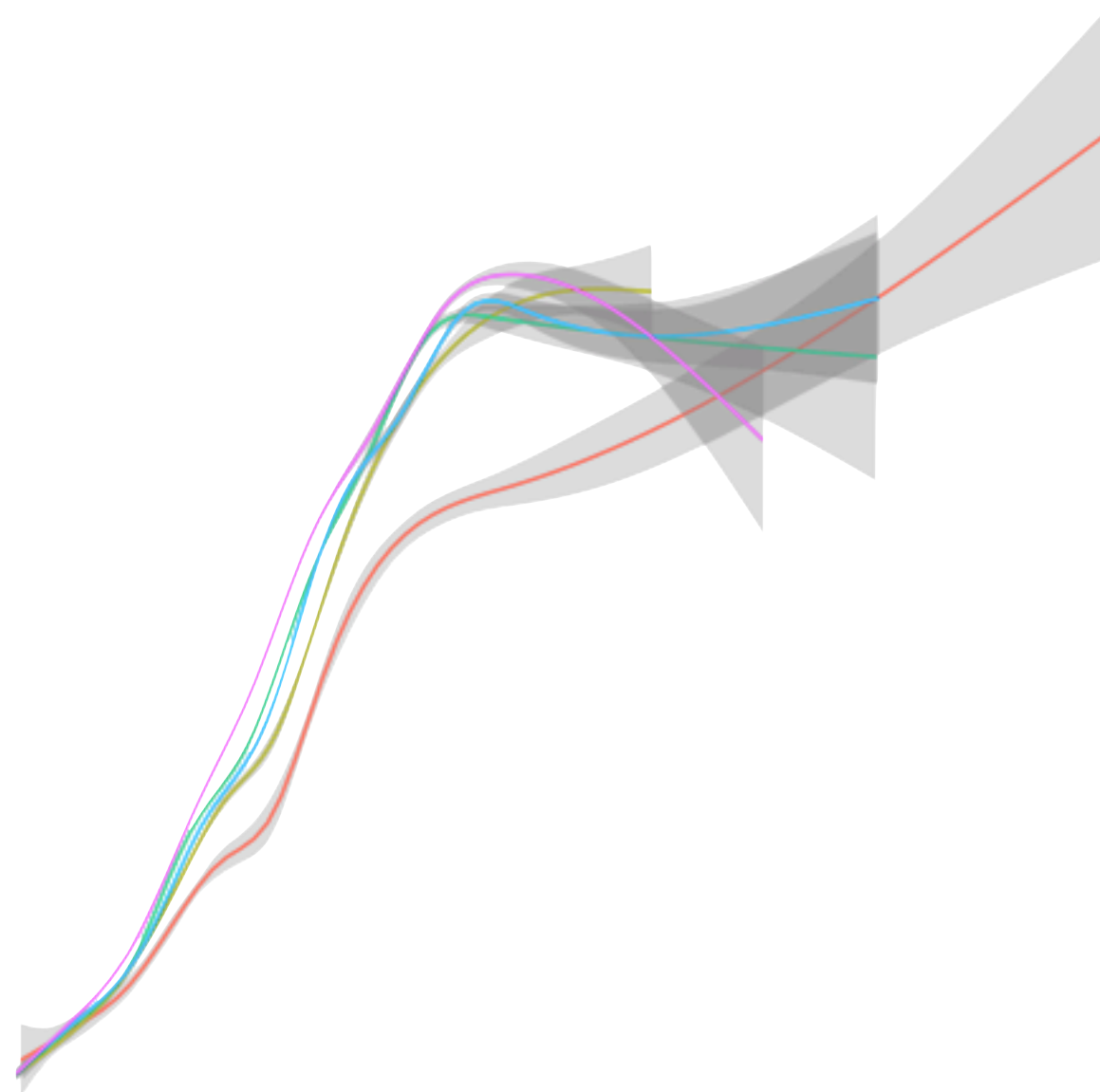
Geom functions provide a way to create separate layers with different aesthetics

What is a layer?



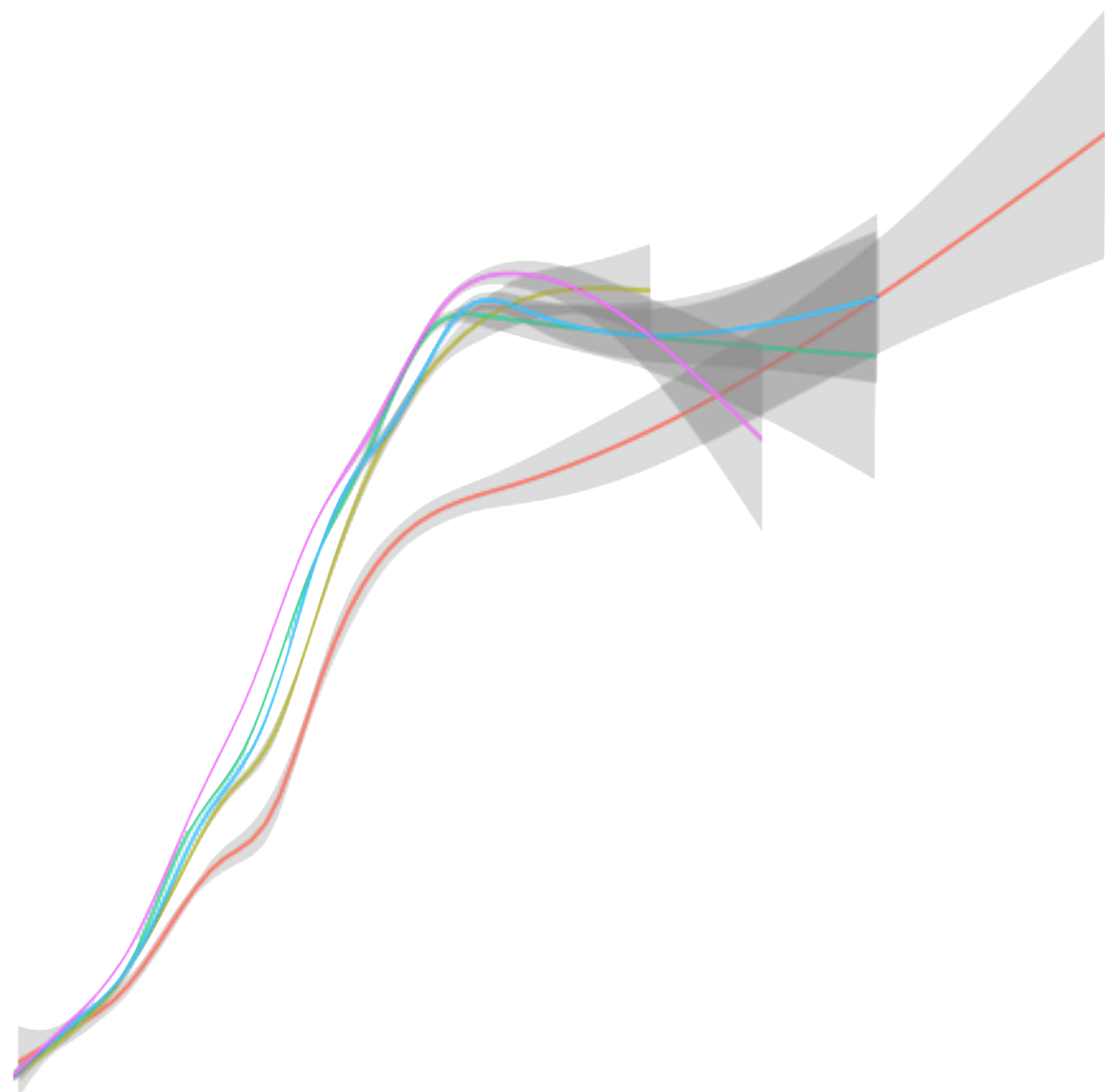
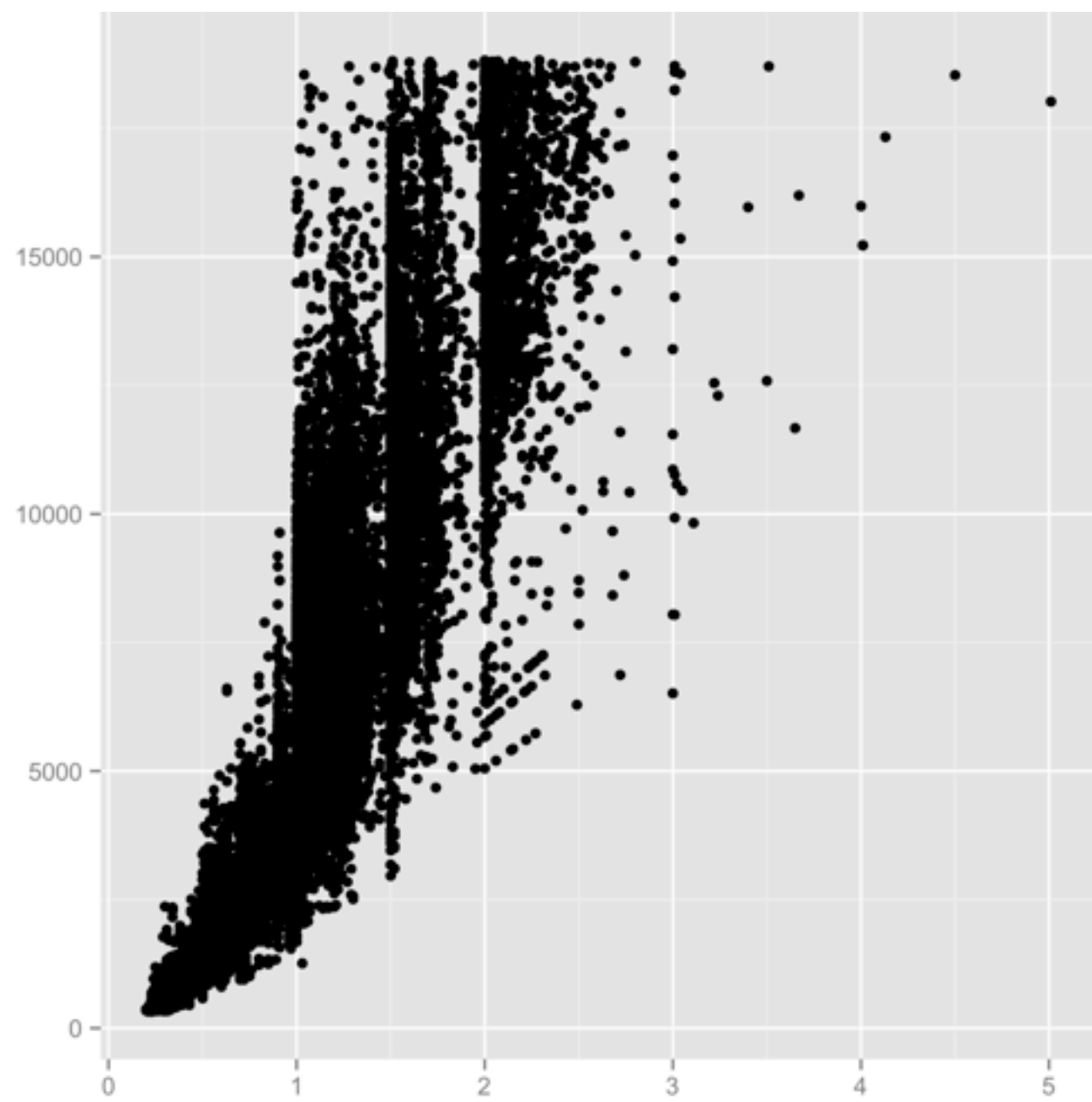


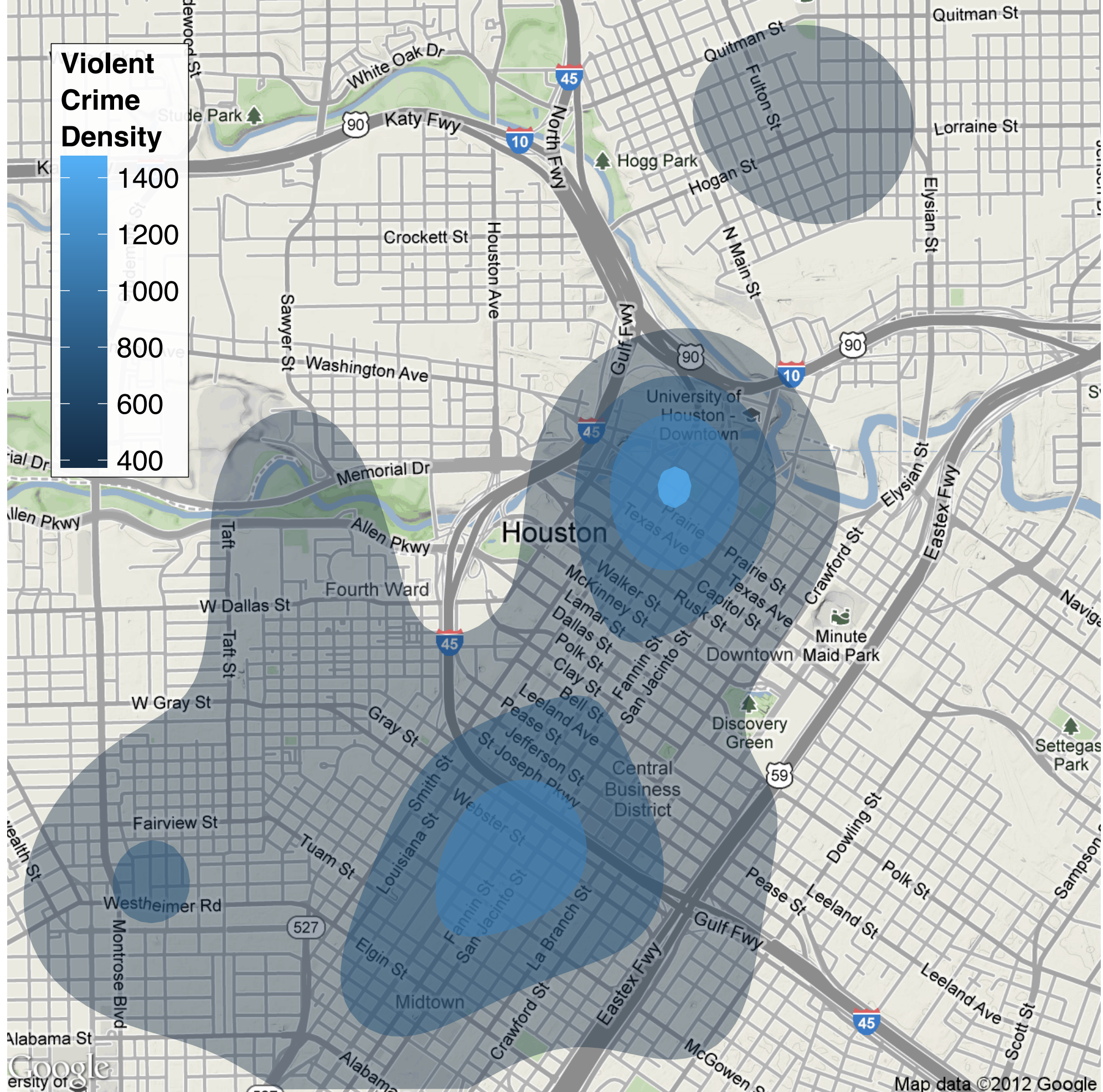
Layer 1



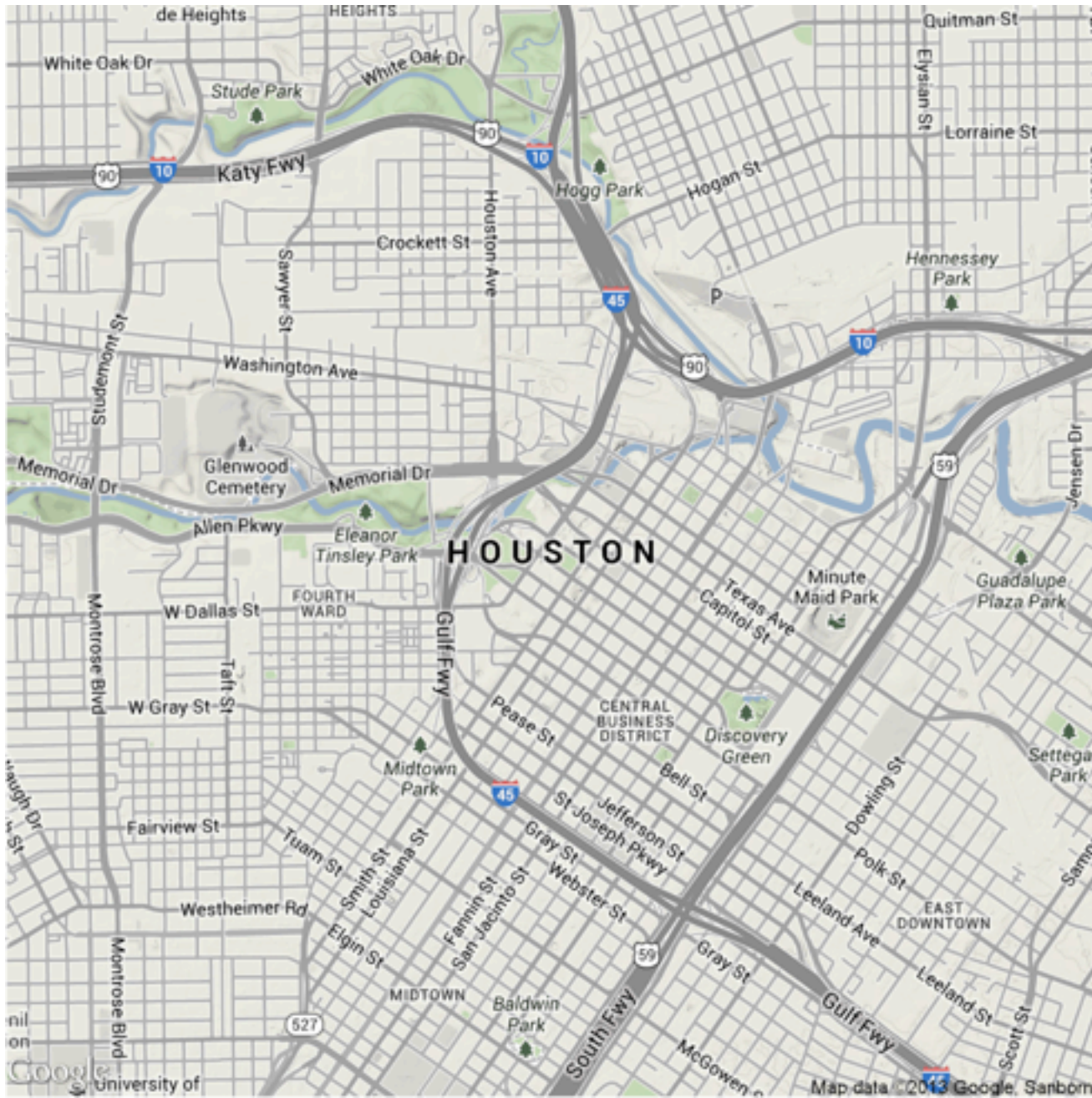
Layer 2



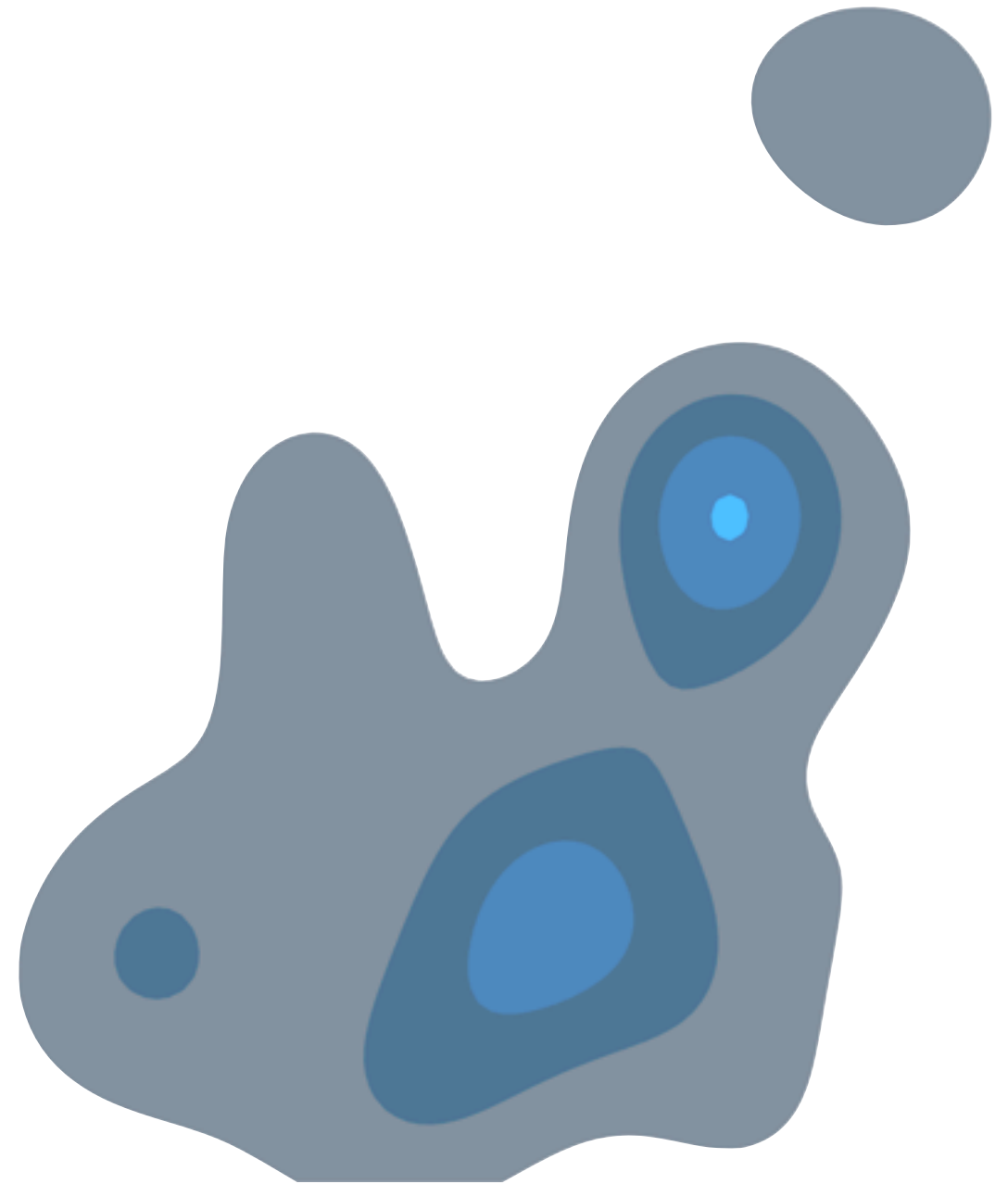




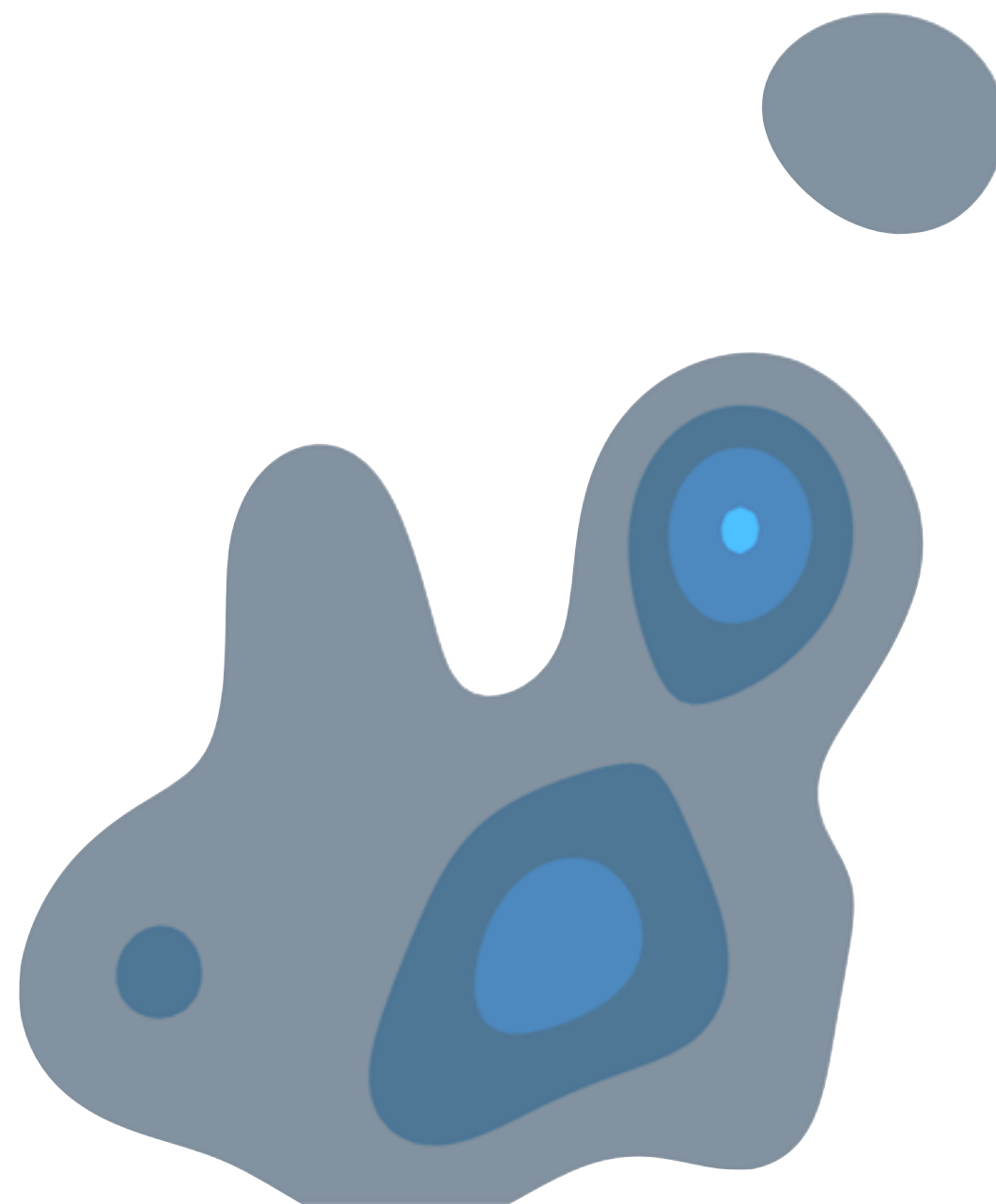
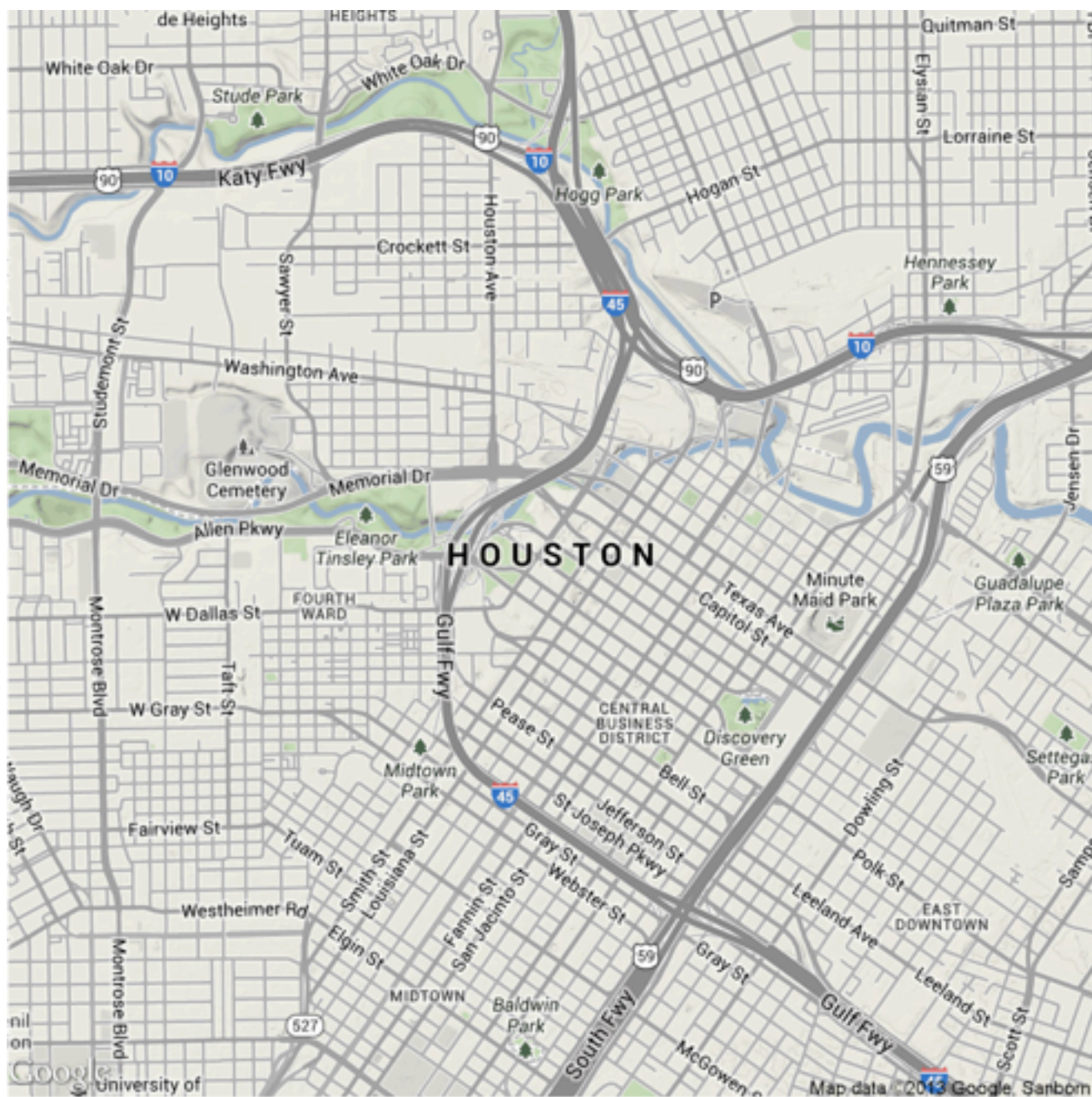




Layer 1



Layer 2





# What is a layer?

Coordinate system

- + geom
- + data
- + aesthetic mappings
- + position adjustment
- + stat



A layer is a  
collection of  
these

# What is a plot?

Coordinate system

+ a layer

+ a layer

+ a layer

+ ...

# ggplot

`ggplot` provides an alternative way to build graphs based on this system.

Its more complicated than `qplot`, but gives you more control.

# ggplot

Coordinate system

+ a layer

+ a layer

+ a layer

+ ...

ggplot() +

# a layer +

# a layer +

# a layer

# ...

But how to build the layers?



# Ever geom is a layer

geom function

aesthetic mapping  
wrapped in the aes function

geom\_smooth(

data set

)

position and stat  
(usually left as the default values)

# ggplot

Coordinate system	<code>ggplot()</code> +
+ a layer	<code>geom_point(</code>
+ a layer	<code>  aes(x = carat, y = price),</code>
+ a layer	<code>  data = diamonds) +</code>
+ ...	<code>geom_smooth(</code>
	<code>  aes(x = carat, y = price,</code>
	<code>      color = cut),</code>
	<code>  data = diamonds)</code>

```
ggplot() +  
  geom_point(aes(x = carat, y = price),  
    data = diamonds) +  
  geom_smooth(aes(x = carat, y = price,  
    color = cut), data = diamonds)
```

A lot of redundant typing

```
ggplot() +  
  geom_point(aes(x = carat, y = price),  
    data = diamonds) +  
  geom_smooth(aes(x = carat, y = price,  
    color = cut), data = diamonds)
```

A lot of redundant typing

```
ggplot() +  
  geom_point(aes(x = carat, y = price),  
    data = diamonds) +  
  geom_smooth(aes(x = carat, y = price,  
    color = cut), data = diamonds)
```

A lot of redundant typing

# set default data and aesthetics with ggplot

ggplot

default  
data set

default aesthetic mappings  
wrapped in the aes function

```
ggplot(diamonds, aes(x = carat, y = price))
```

You can overwrite or add to  
the defaults at the layer level

ggplot

data

global aesthetics

```
ggplot(diamonds, aes(x = carat, y = price)) +  
  geom_point() +  
  geom_smooth(aes(color = cut))
```

layer

layer specific aesthetics

# Your turn

Use ggplot to make these graphs.

ggplot

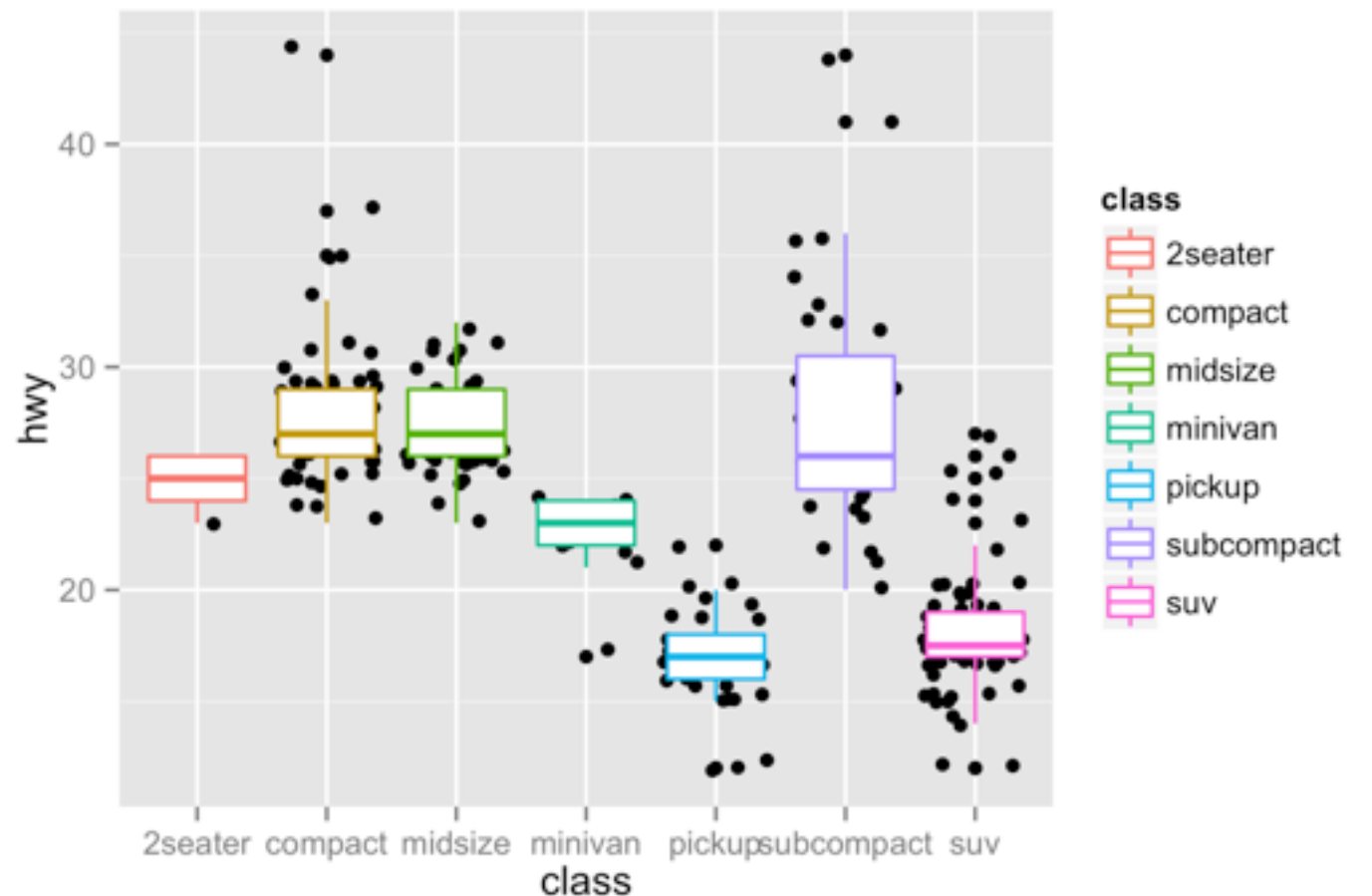
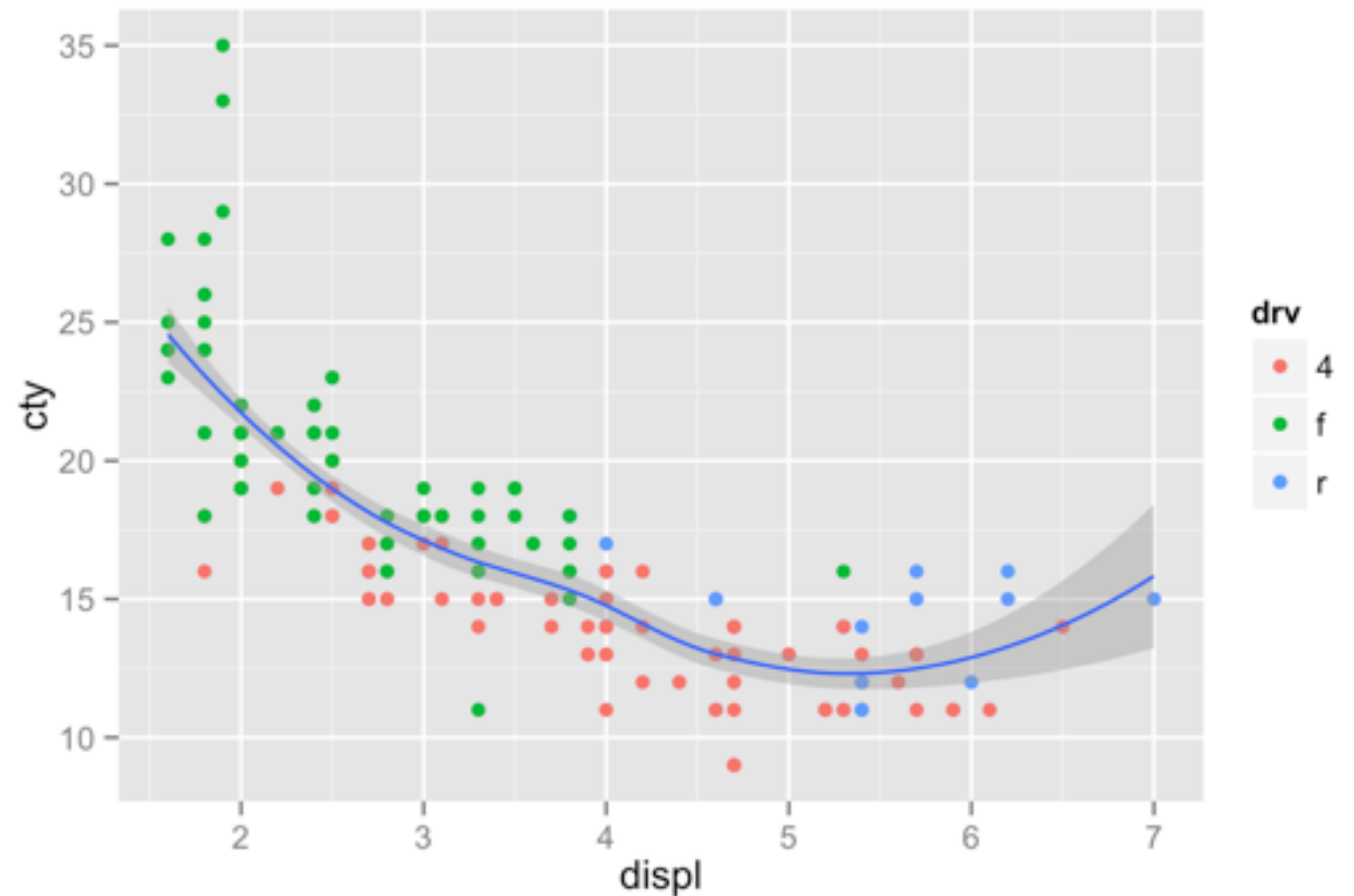
data

global aesthetics

```
ggplot(diamonds, aes(x=carat, y=price)) +  
  geom_point() +  
  geom_smooth(aes(color = cut))
```

layer

layer specific aesthetics





```
ggplot(mpg, aes(displ, cty)) +  
  geom_point(aes(color = drv)) +  
  geom_smooth()
```

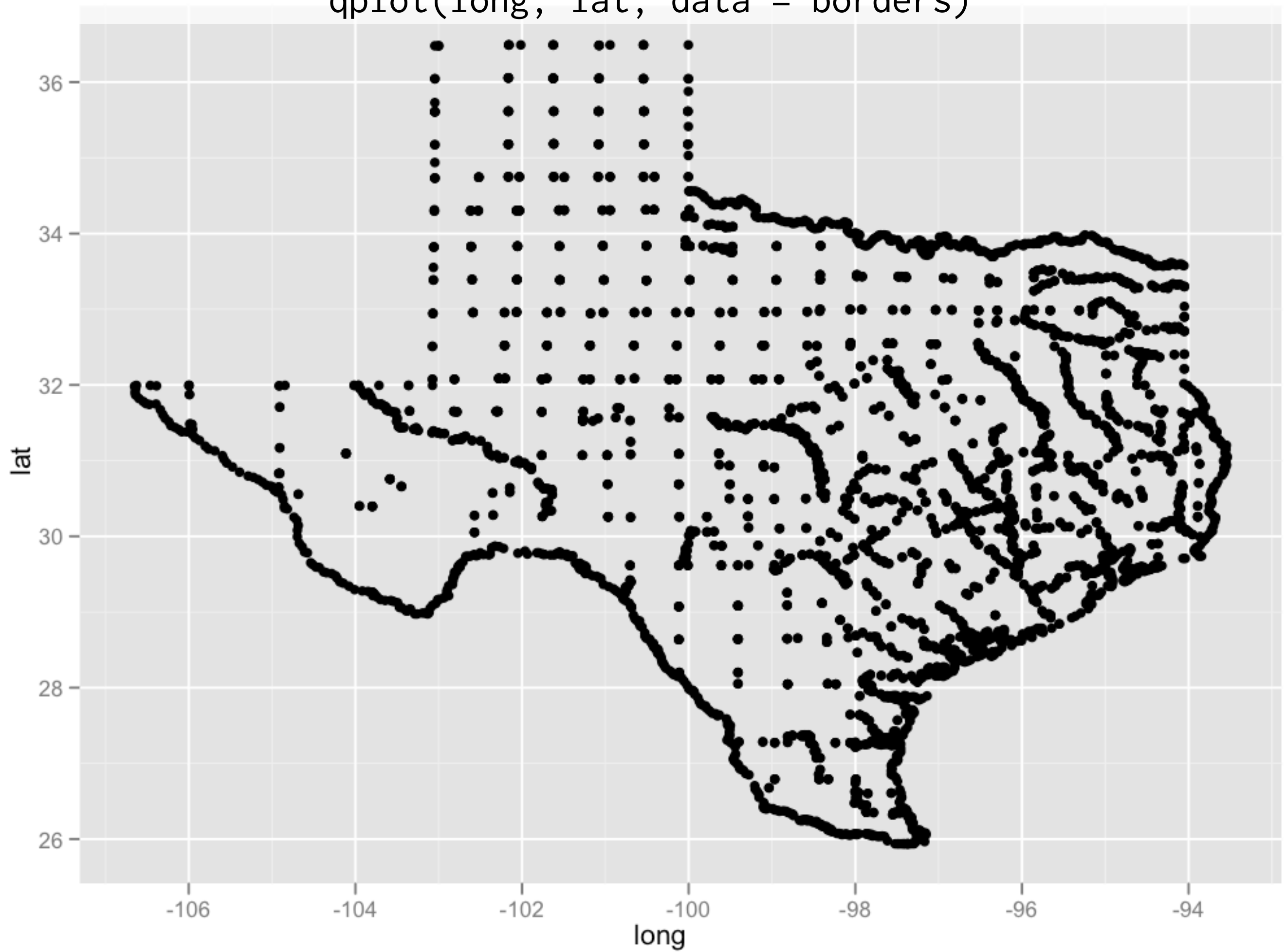
```
ggplot(mpg, aes(class, hwy)) +  
  geom_point(position = "jitter") +  
  geom_boxplot(aes(color = class))
```

# **Customizing graphics**

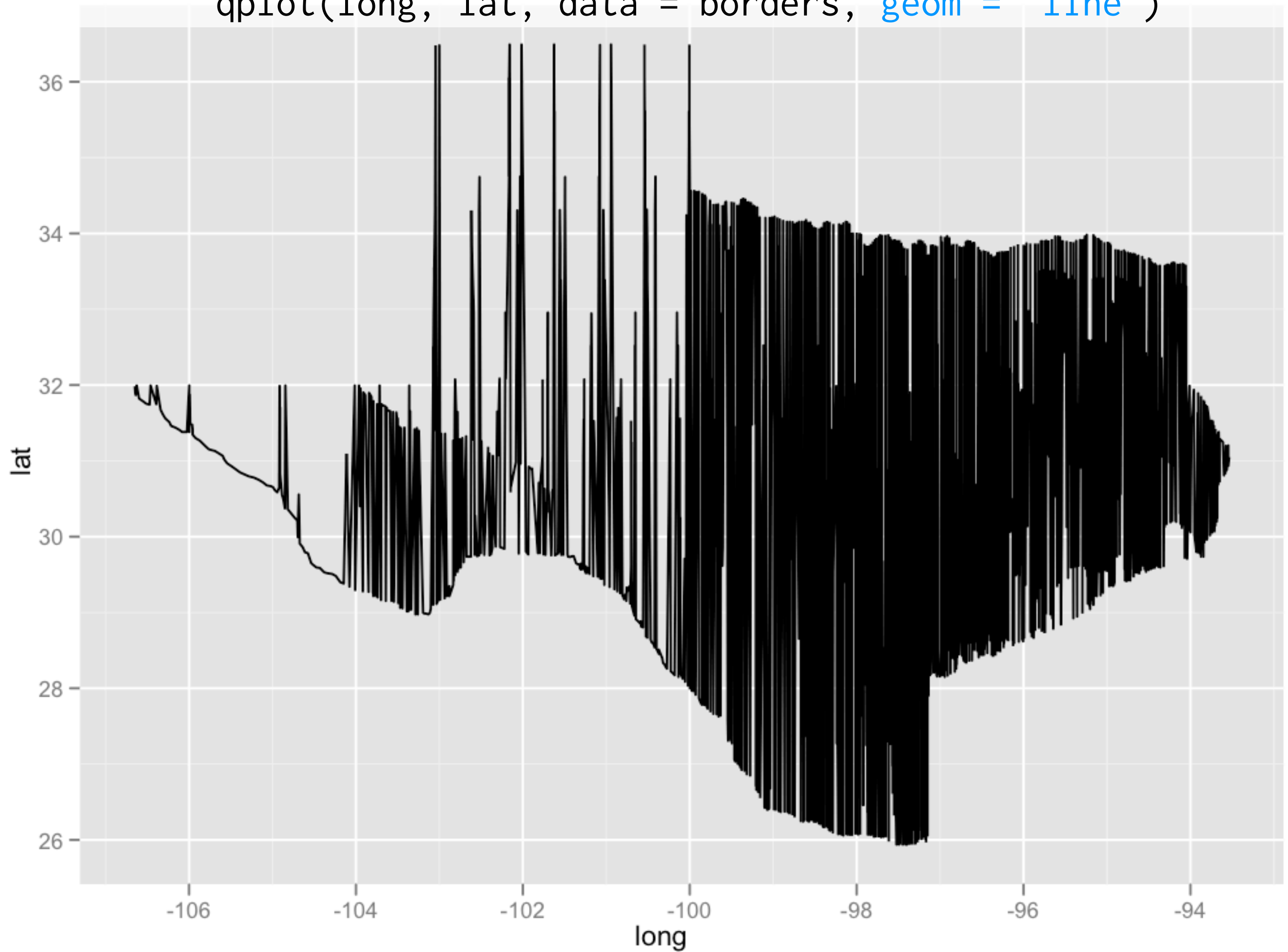
# Texas population data

```
borders <- read.csv("data/texas.csv")  
View(borders)
```

```
qplot(long, lat, data = borders)
```



```
qplot(long, lat, data = borders, geom = "line")
```



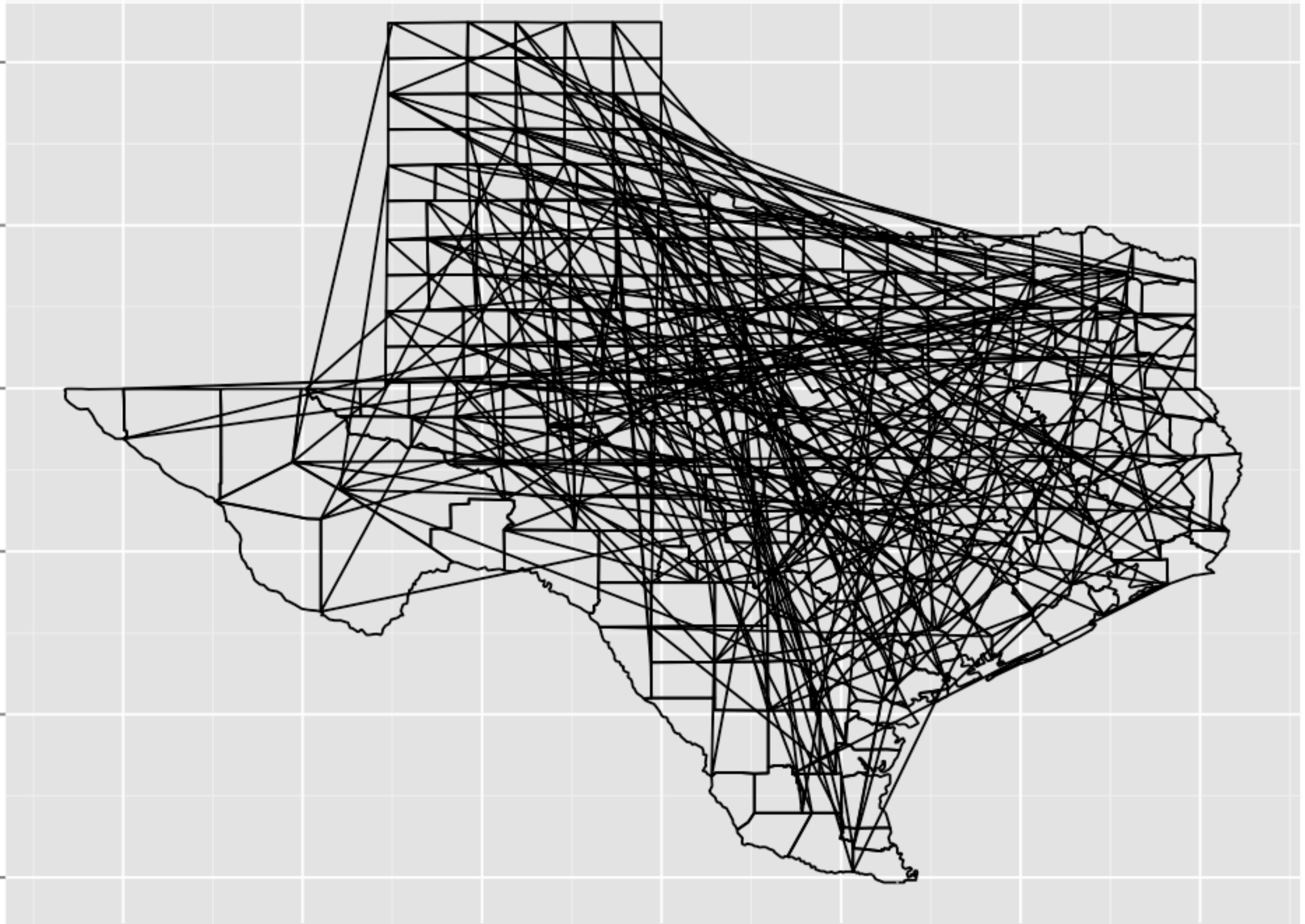
```
qplot(long, lat, data = borders, geom = "path")
```

lat

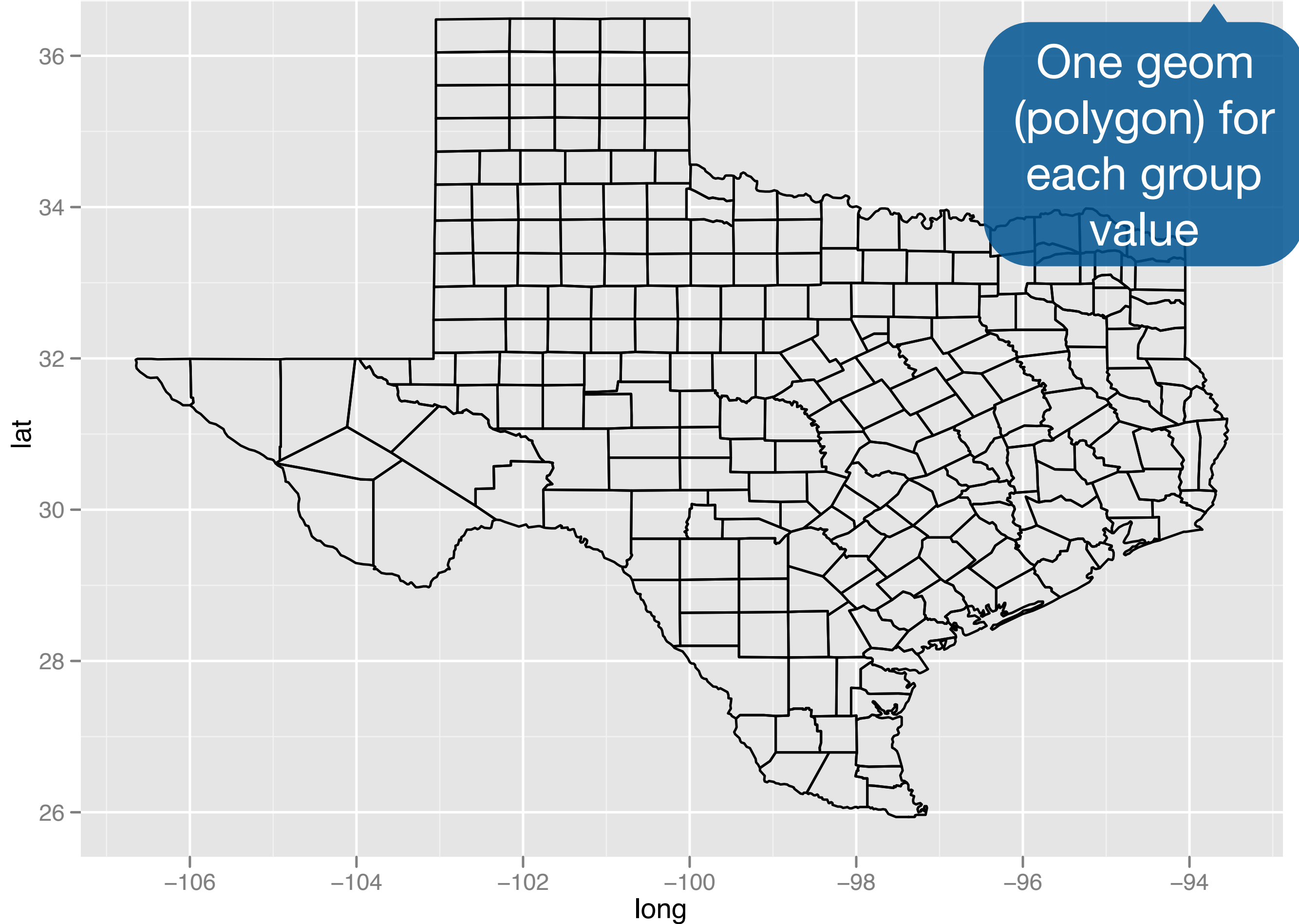
36  
34  
32  
30  
28  
26

long

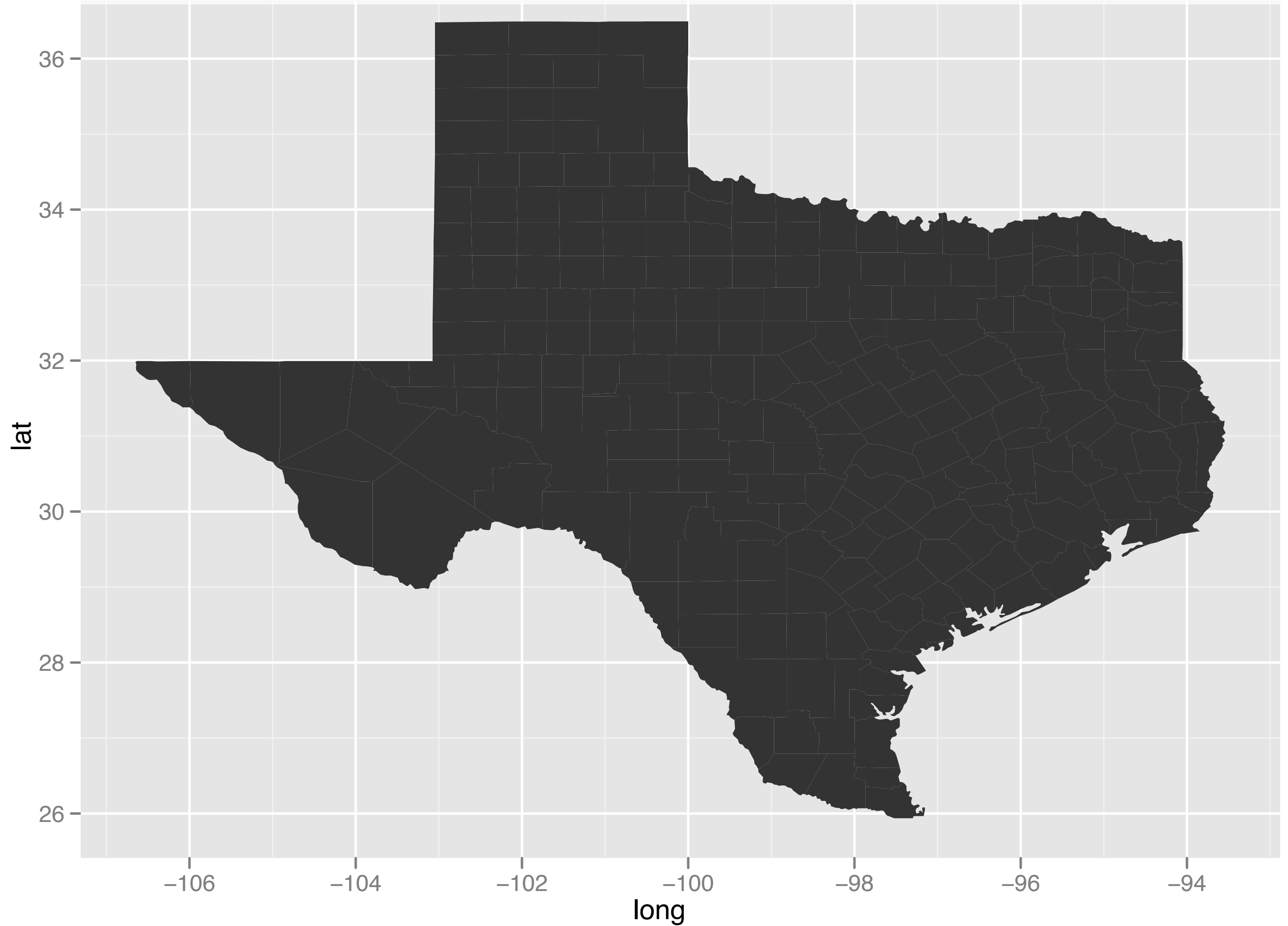
-106 -104 -102 -100 -98 -96 -94



```
qplot(long, lat, data = borders, geom = "path", group = group)
```



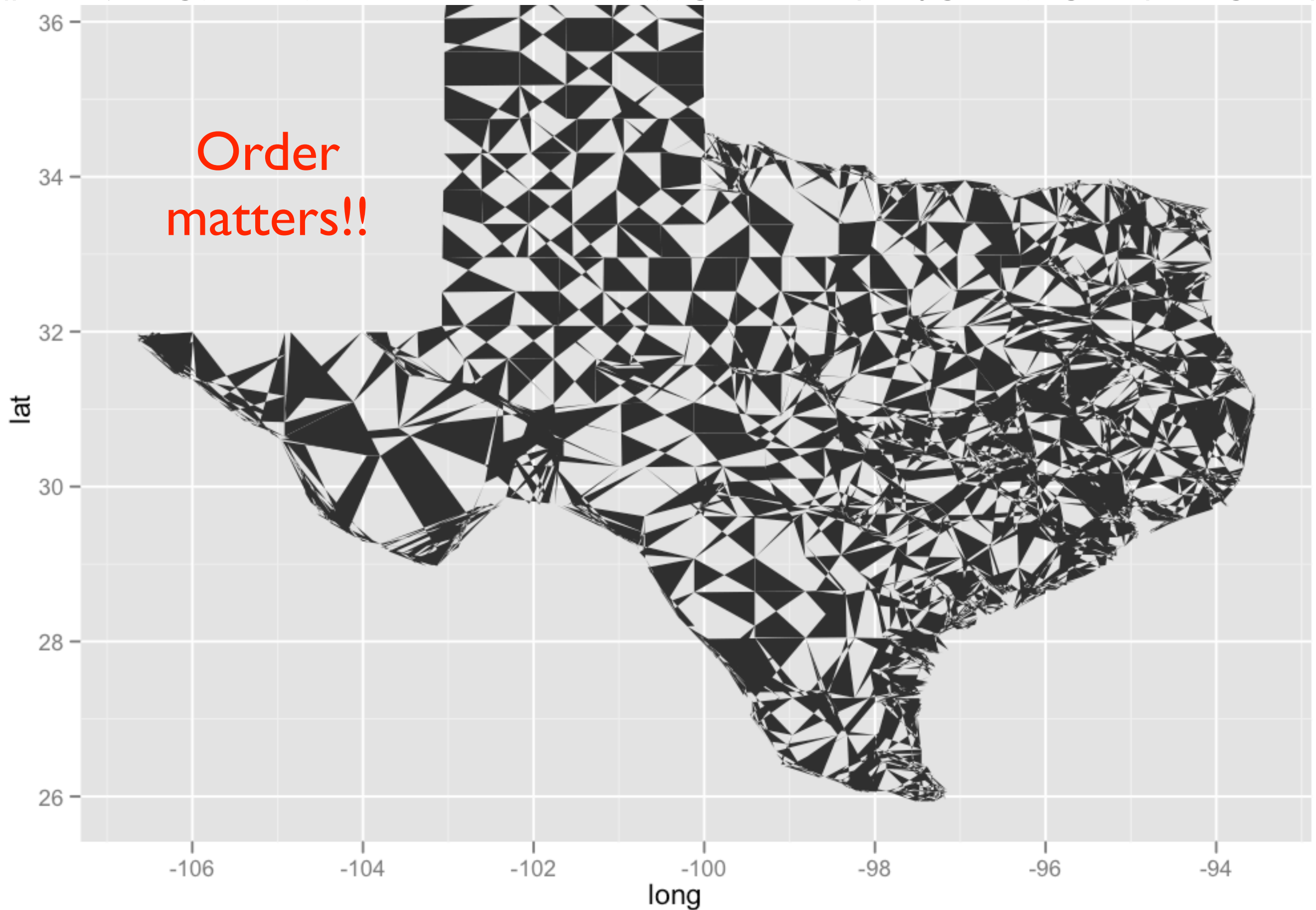
```
qplot(long, lat, data = borders, geom = "polygon", group = group)
```



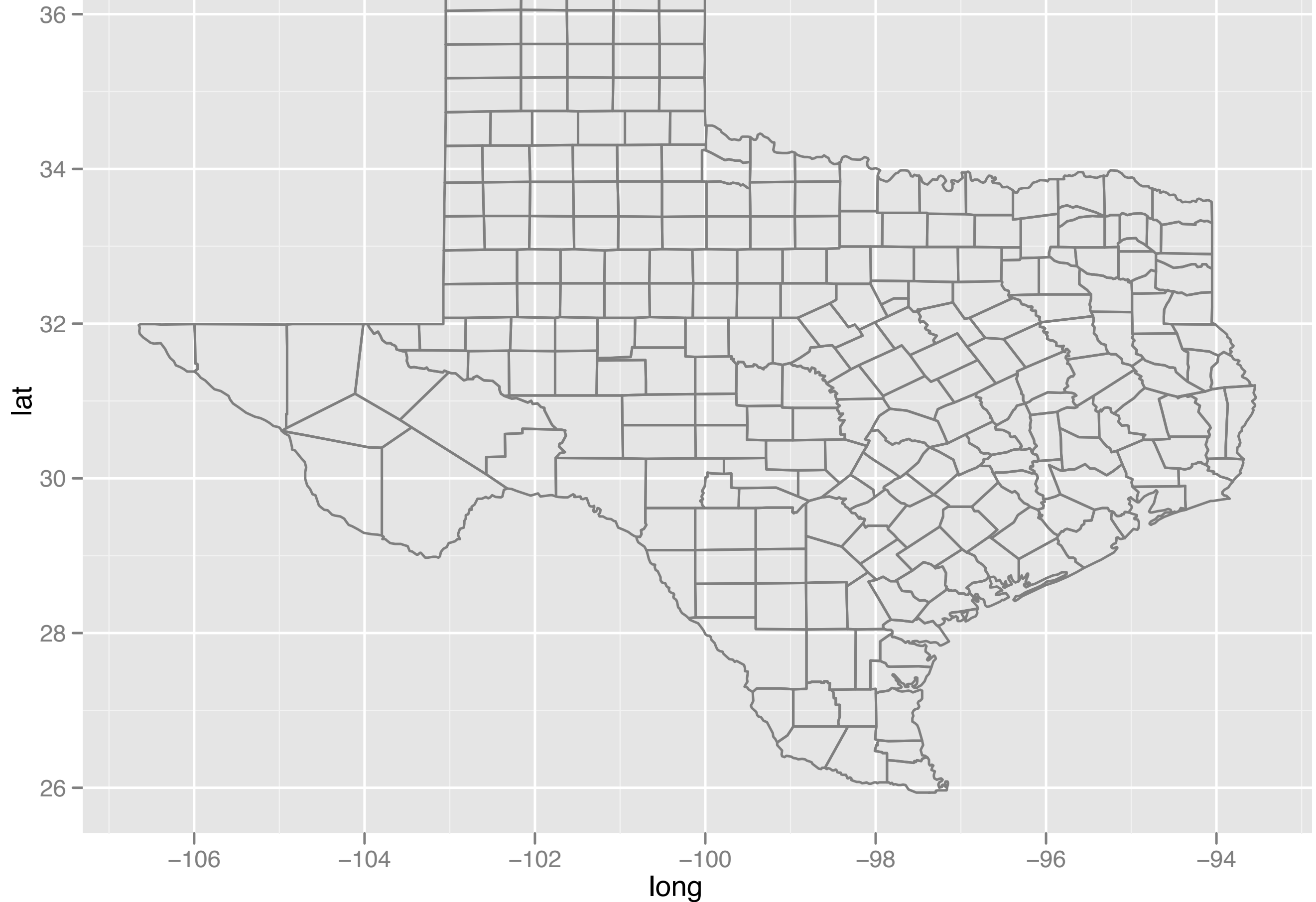


```
borders2 <- borders[sample(nrow(borders)), ]  
qplot(long, lat, data = borders2, geom = "polygon", group = group)
```

Order  
matters!!

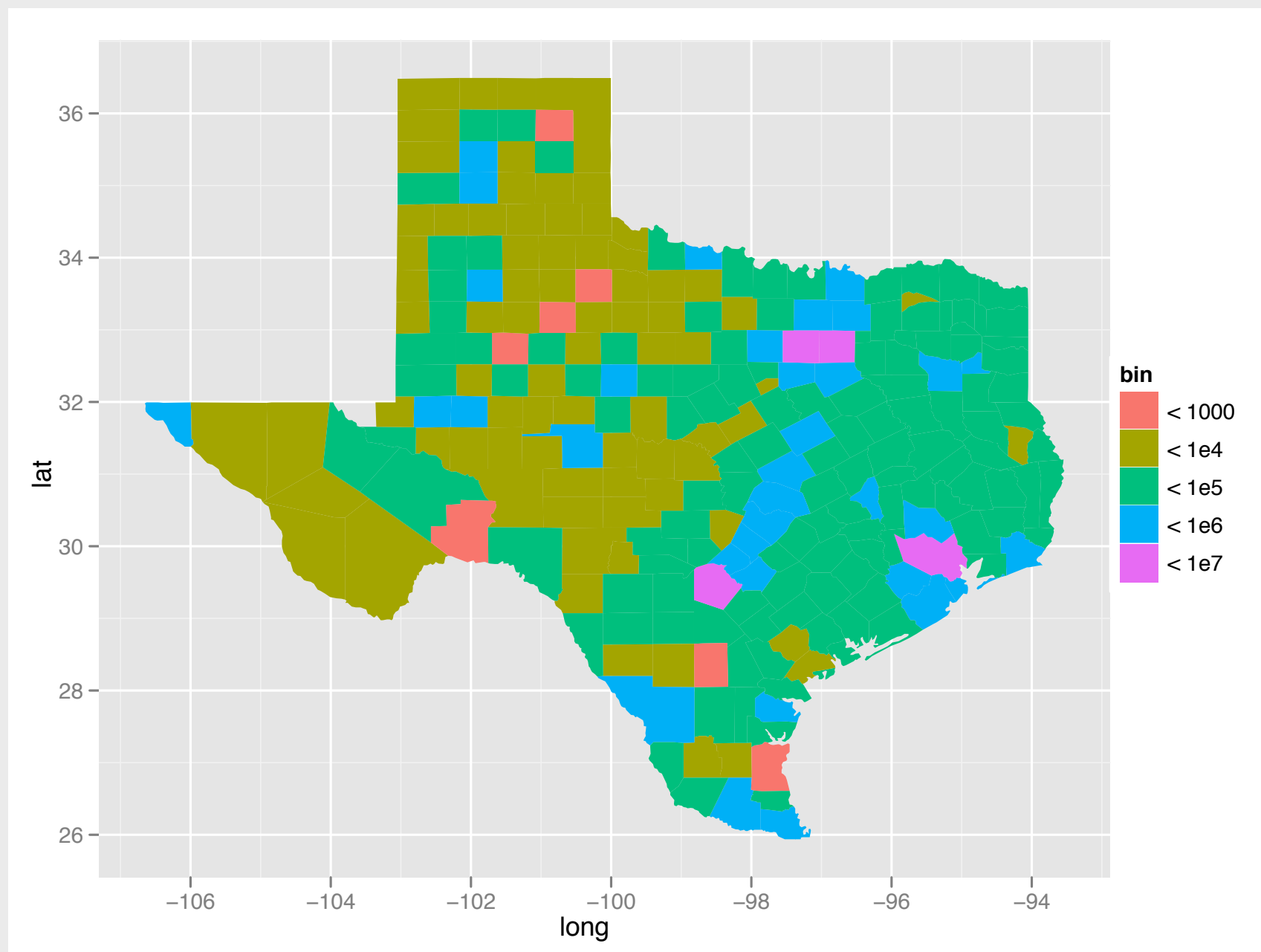


```
ggplot(borders, aes(long, lat)) +  
  geom_polygon(aes(group = group), fill = NA, color = "grey50")
```



# Your Turn

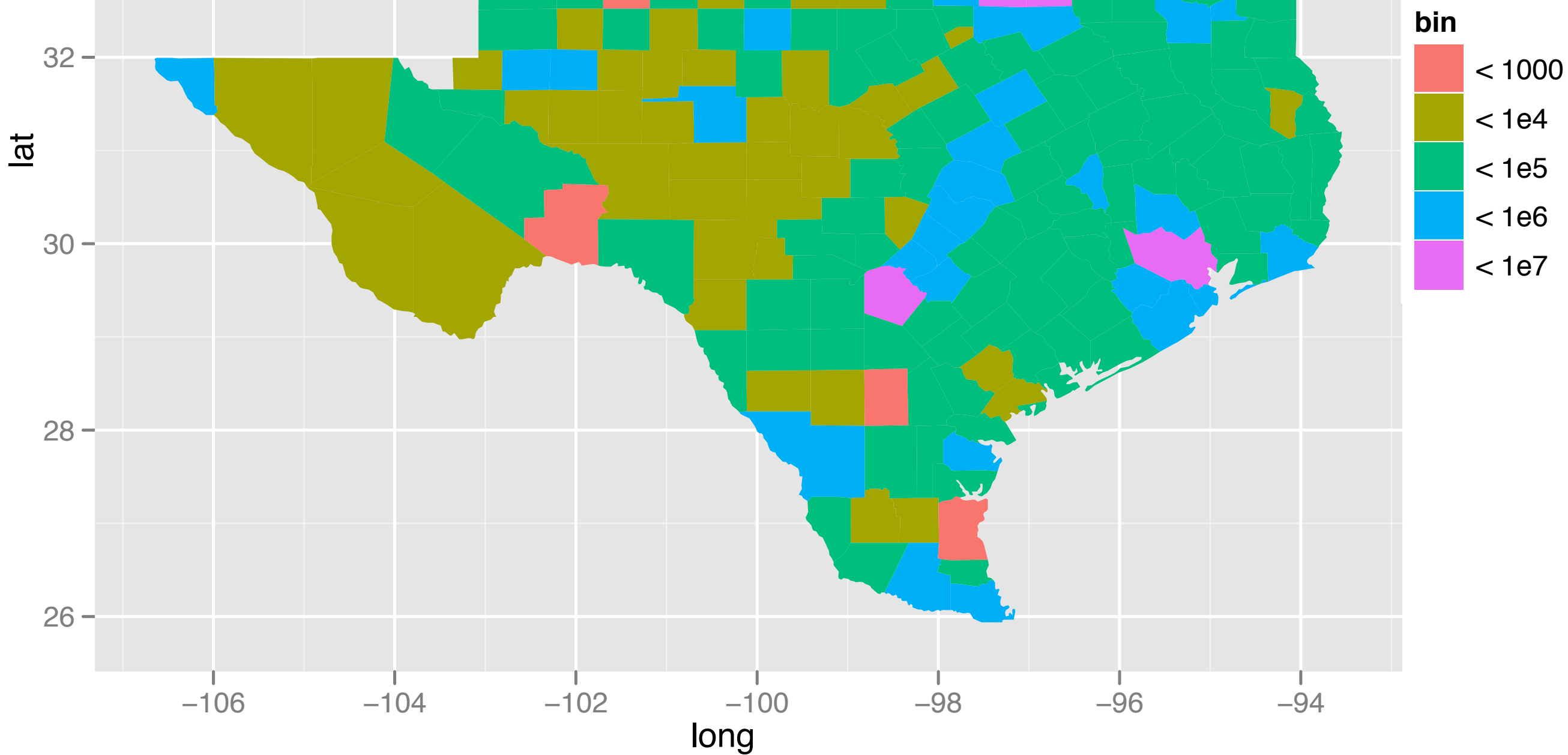
Use borders to recreate this map.



```
tx <- qplot(long, lat, data = borders, geom =  
"polygon", fill = bin, group = group)
```

```
tx
```

**What's  
wrong  
with this  
plot?**



# Some problems

Incorrect aspect ratio

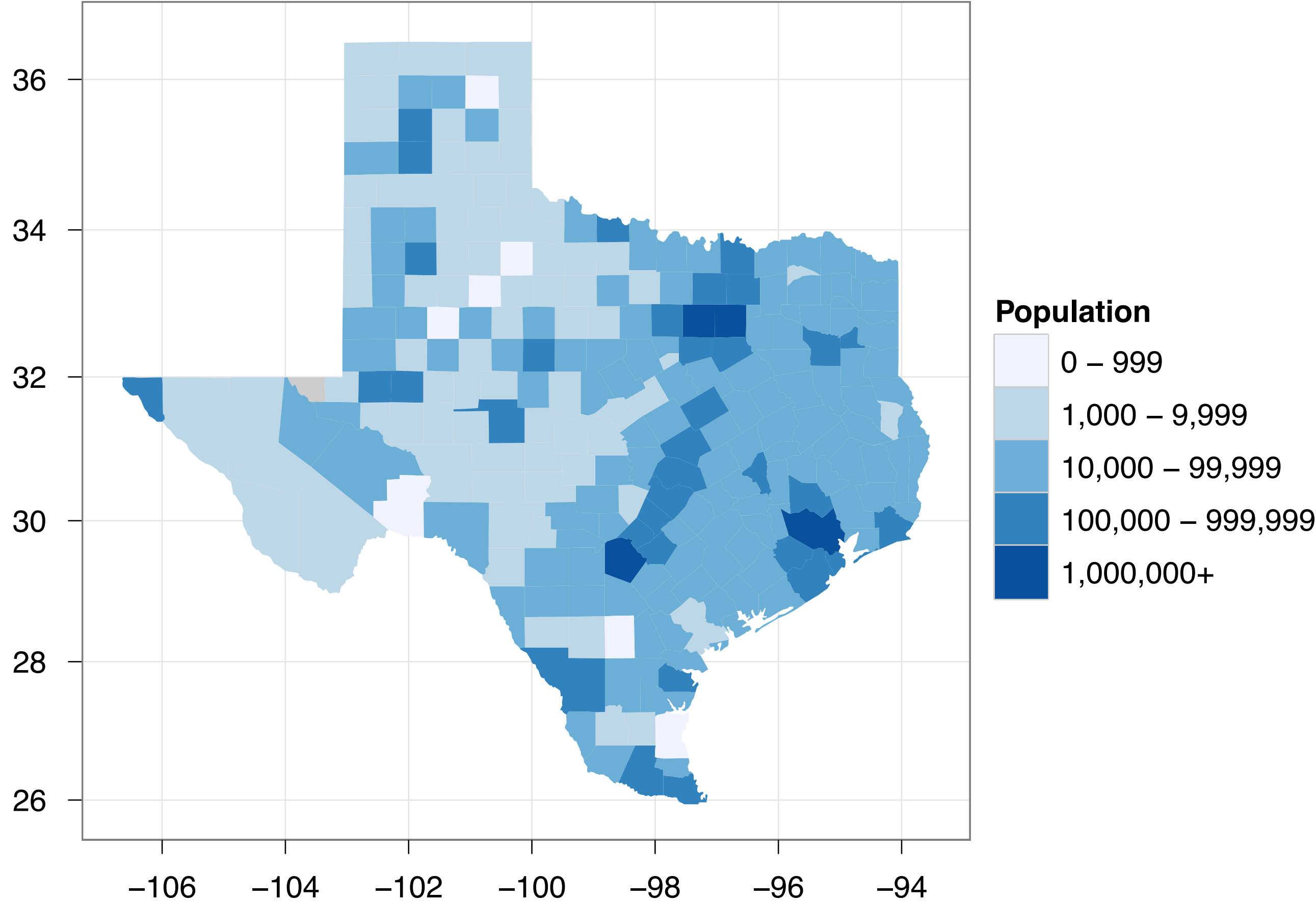
Bad color scheme

Unnecessary axis labels

Legend needs improvement: better title  
and better key labels

No title

# Population of Texas Counties



**Title**



# ggtitle

```
tx + ggtitle("Population of Texas Counties")
```

# ggplot2 syntax


You modify ggplot2 graphs by adding objects to them.

```
tx + ggtitle("Population of Texas Counties")
```

# ggplot2 syntax

You modify ggplot2 graphs by adding objects to them.

```
tx + ggtitle("Population of Texas Counties")
```



Creates a  
ggplot2 title

# ggplot2 syntax

You modify ggplot2 graphs by adding objects to them.

```
tx + ggtitle("Population of Texas Counties")
```

Adds it to the  
graph

Creates a  
ggplot2 title

Additions are not permanent. They just affect the current graph being drawn

```
tx + ggtitle("Population of Texas Counties")
```

```
tx
```

# to create a new graph that always has a title

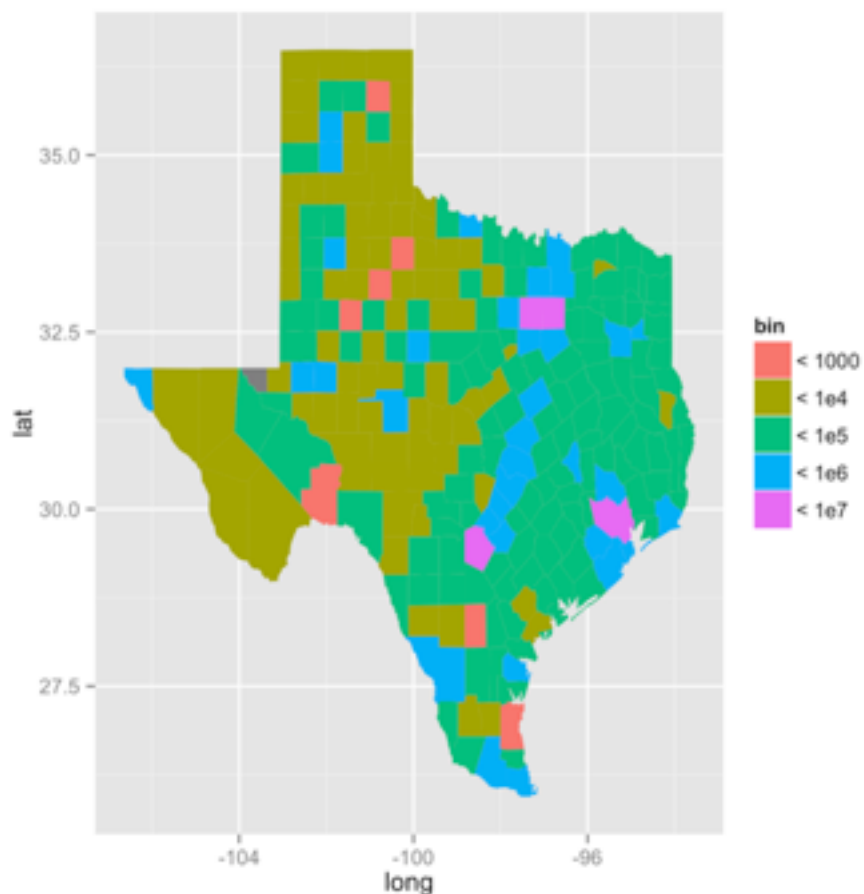
```
tx2 <- tx + ggtitle("Population of Texas Counties")
```

```
tx2
```

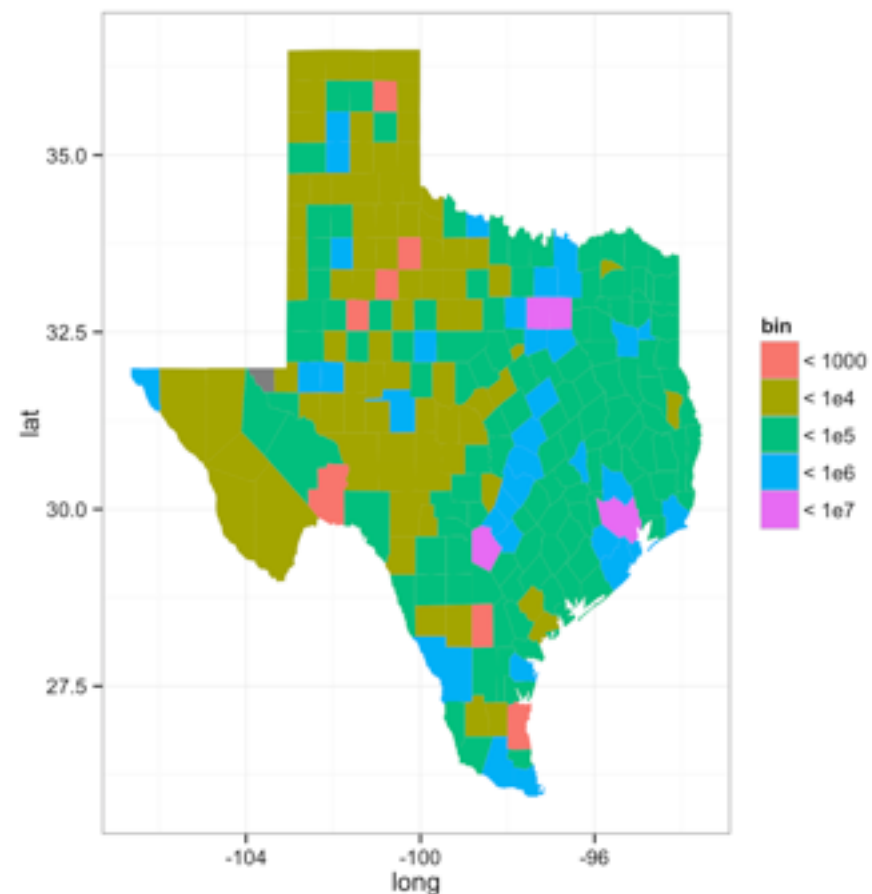
# Themes

# themes

ggplot2 comes with two pre-loaded themes that control the appearance of non-data elements



`tx + theme_grey()`



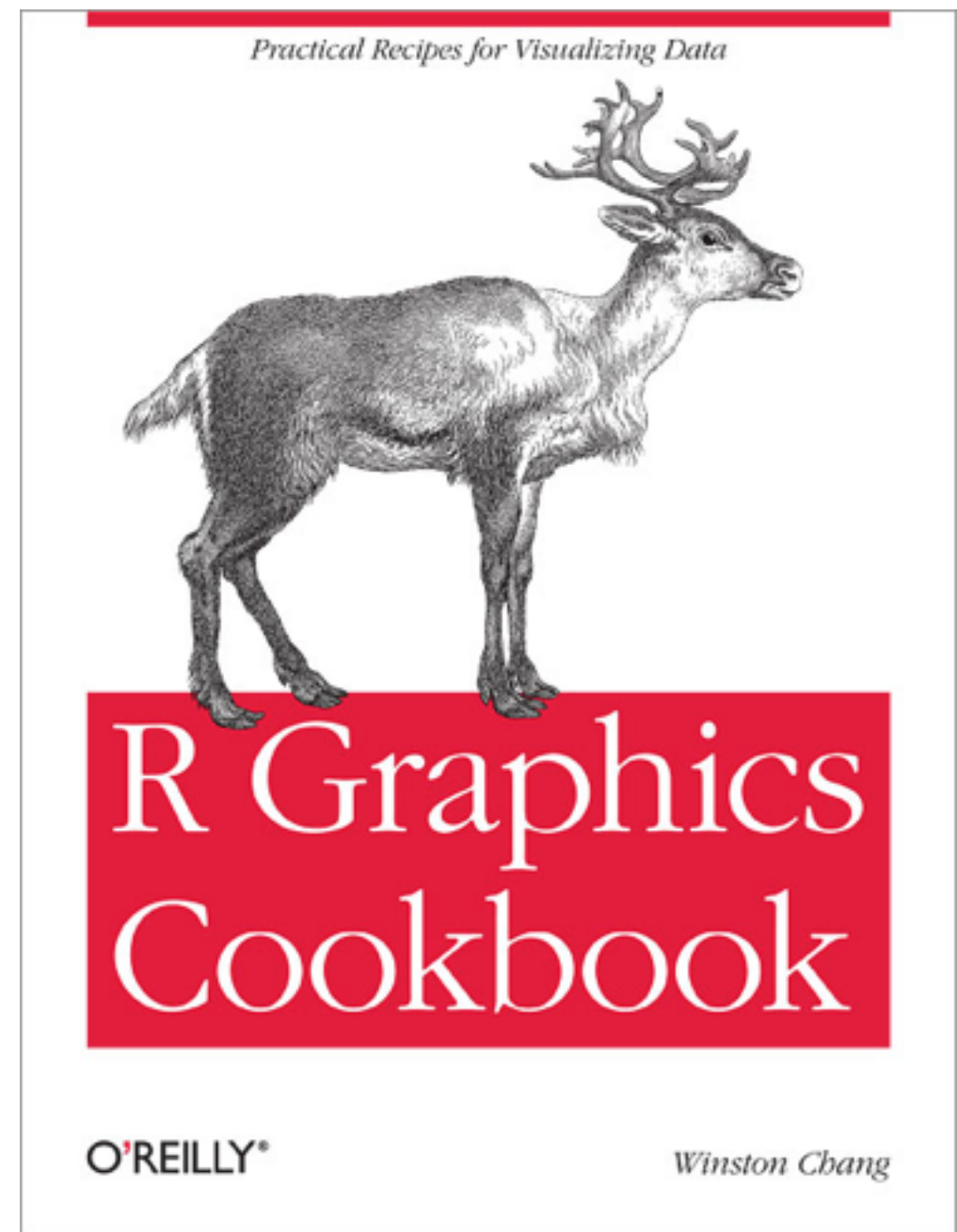
`tx + theme_bw()`

# themes

The ggthemes package offers other pre-built themes.

To learn how to change individual elements of a theme, I recommend the R Graphics Cookbook by Winston Chang

<http://shop.oreilly.com/product/0636920023135.do#>





# Coordinate systems

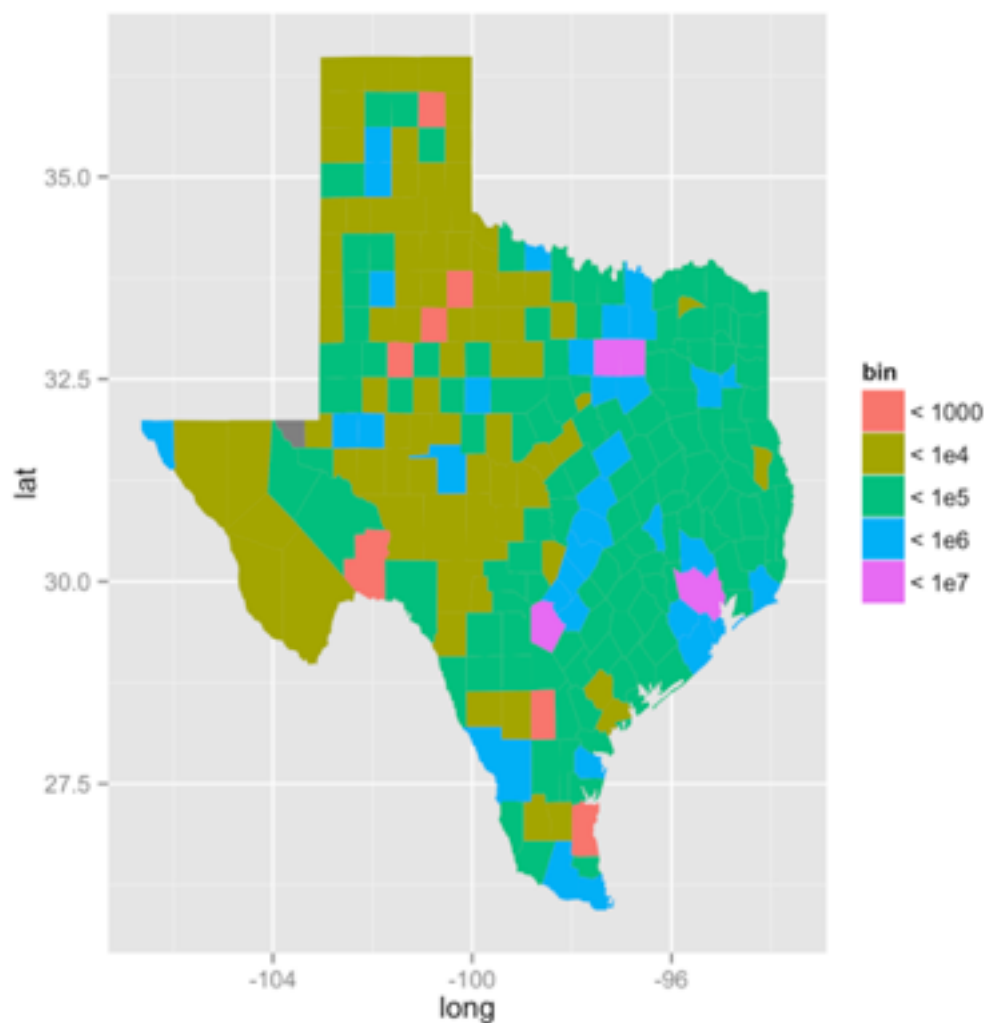
# coordinate systems

ggplot2 comes with a few different coordinate systems, but you'll almost always use cartesian coordinates

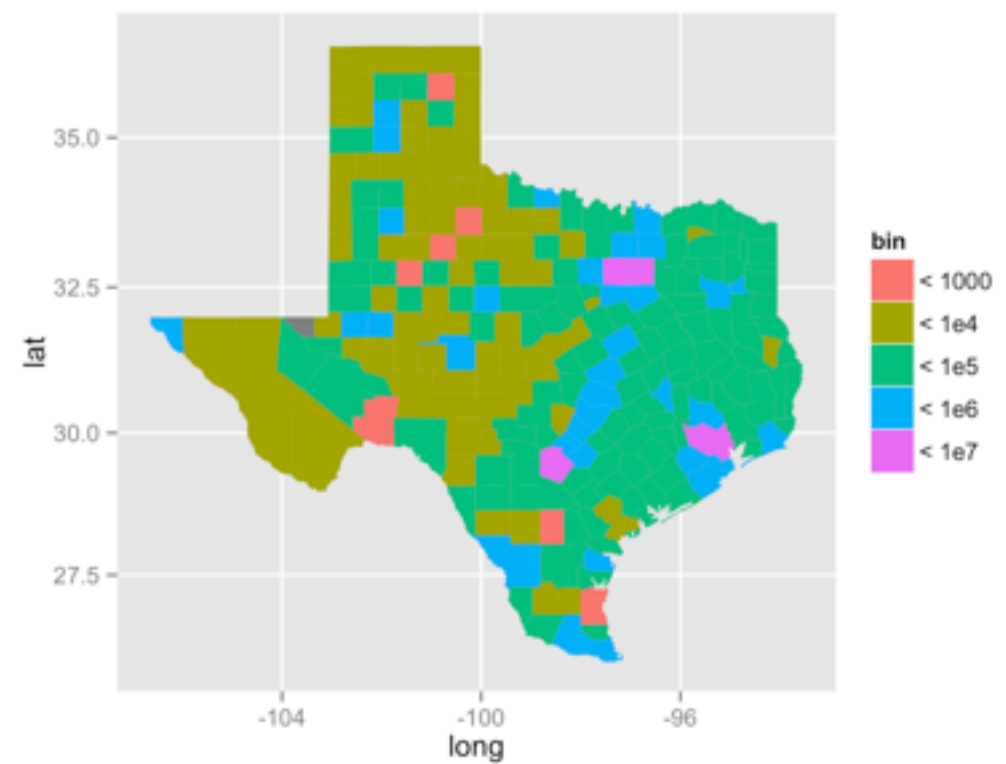
function	system
<code>coord_cartesian()</code>	Cartesian coordinates
<code>coord_fixed()</code>	Cartesian with a fixed aspect ratio
<code>coord_polar()</code>	Polar coordinates
<code>coord_map()</code>	A map projection

# coordinates

Default



```
tx + coord_cartesian()
```



```
tx + coord_map("mercator")
```

# Your Turn

Modify tx to

1. have a title
2. use the black and white theme
3. use a mercator map projection

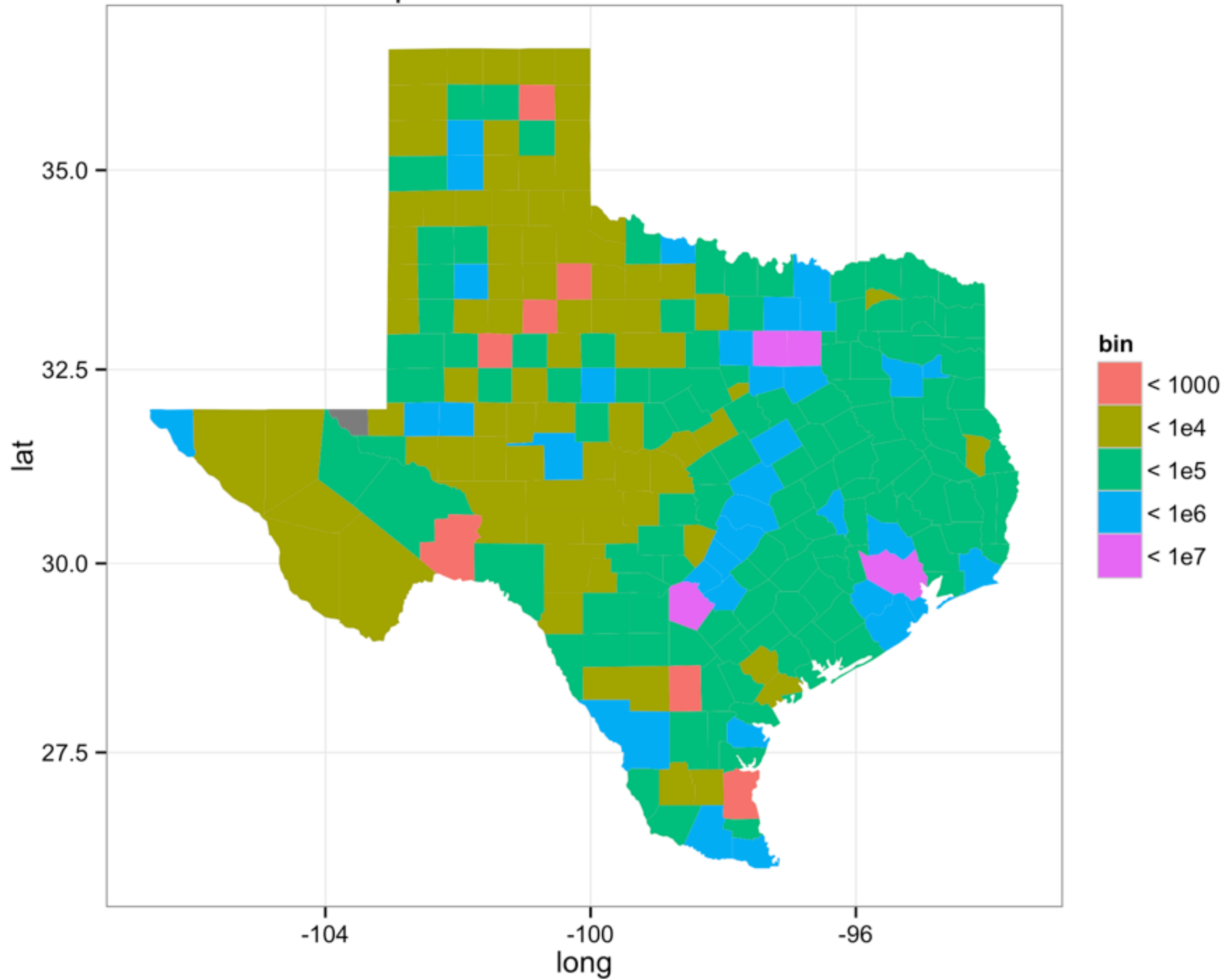
```
tx +
```

```
  ggtitle("Population of Texas Counties") +
```

```
  theme_bw() +
```

```
  coord_map("mercator")
```

# Population of Texas Counties

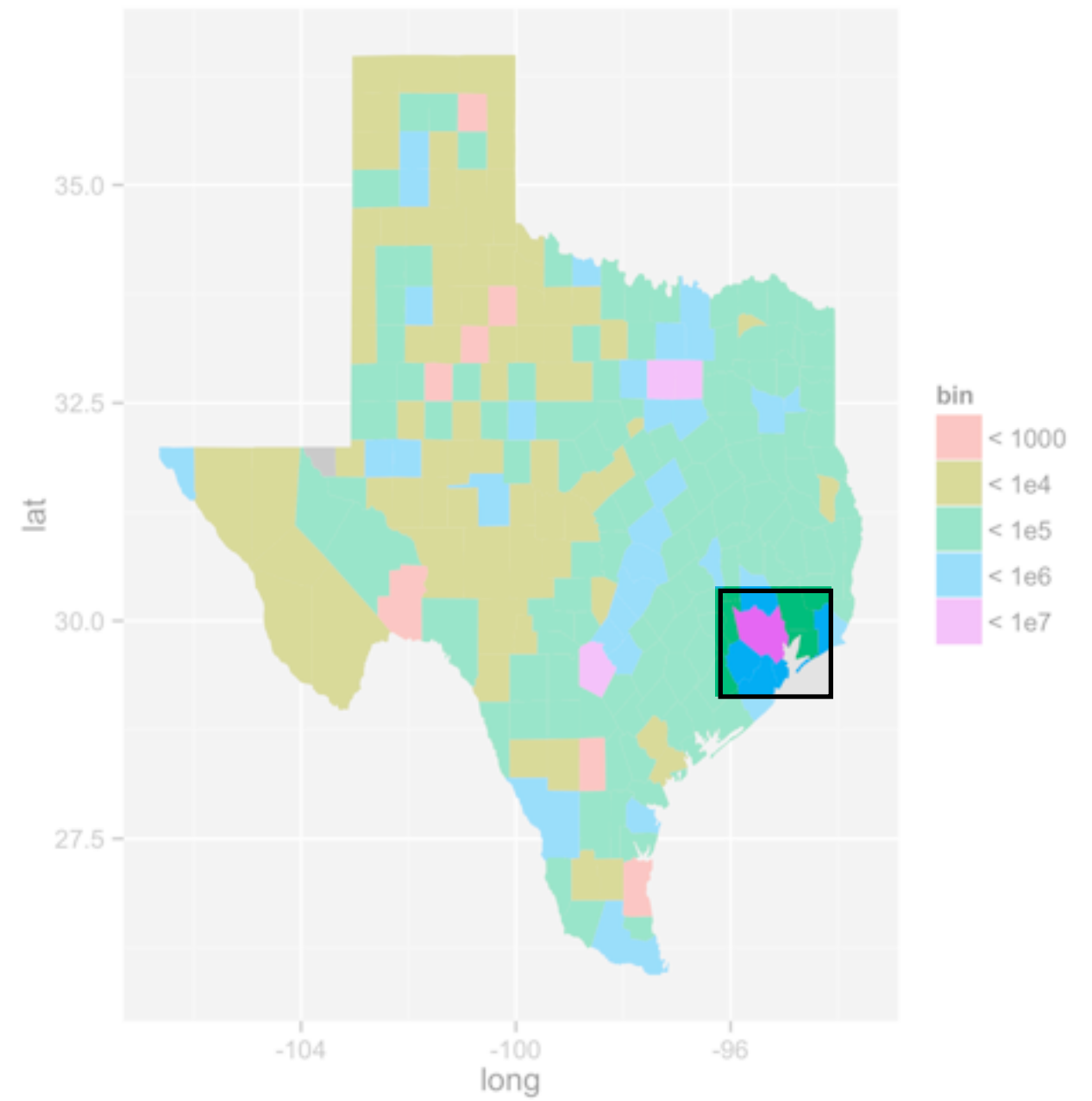
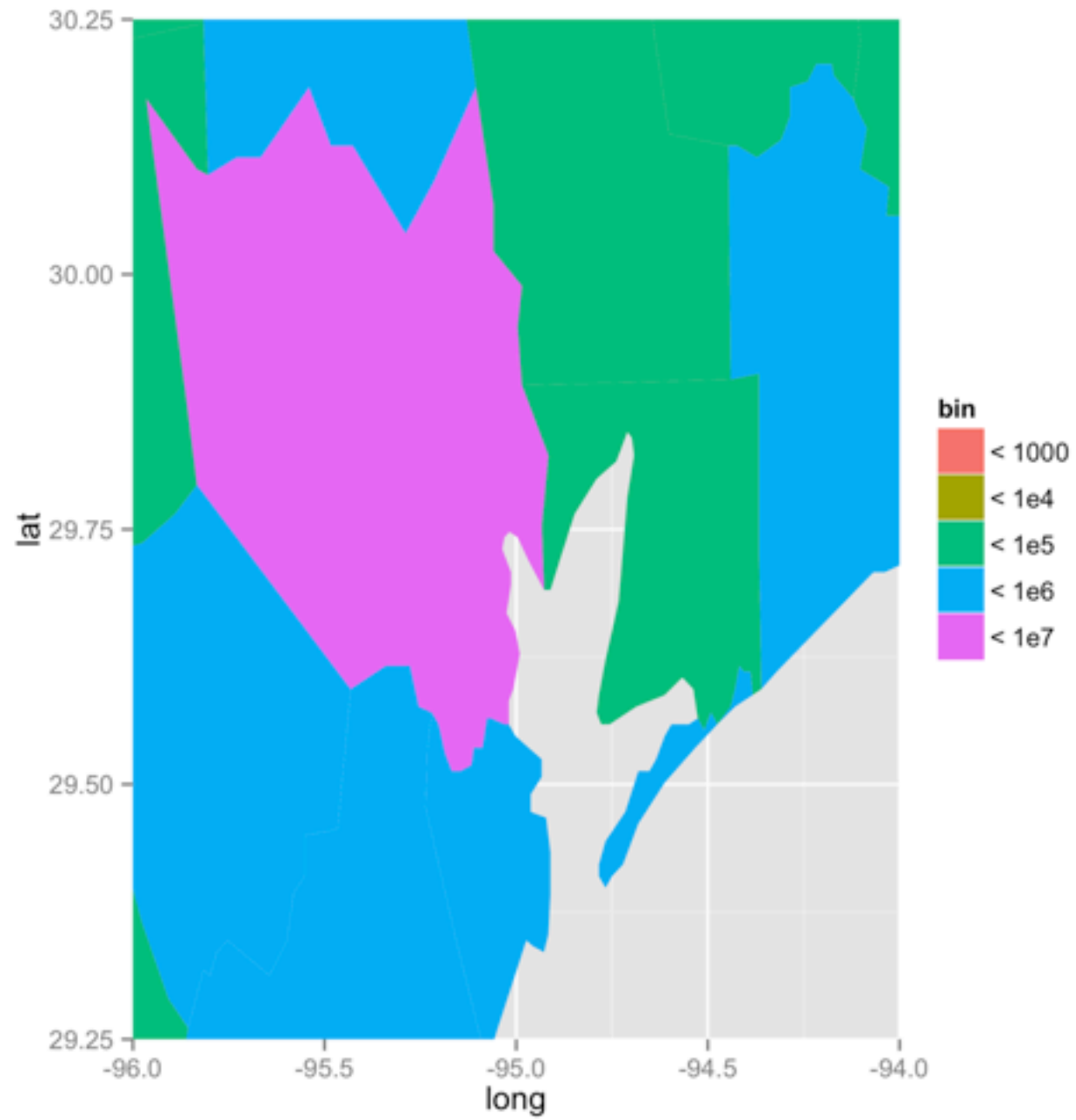


# Aside: zooming

The coordinate system provides a convenient way to zoom

```
# Zoomed in on Houston, Texas
```

```
tx + coord_cartesian(xlim = c(-96, -94),  
  ylim = c(29.25, 30.25))
```





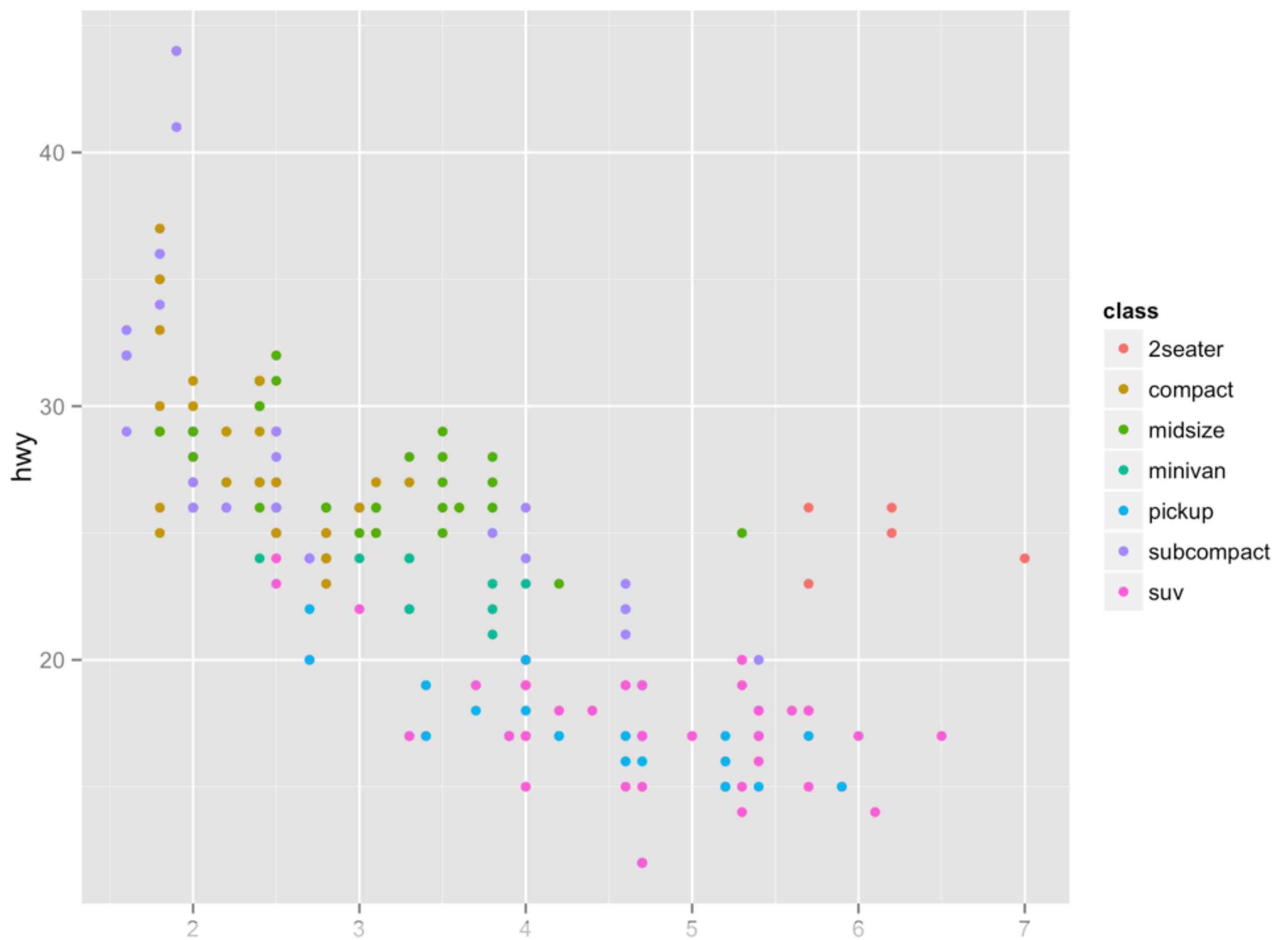
# Scales

# **Aesthetic mapping**

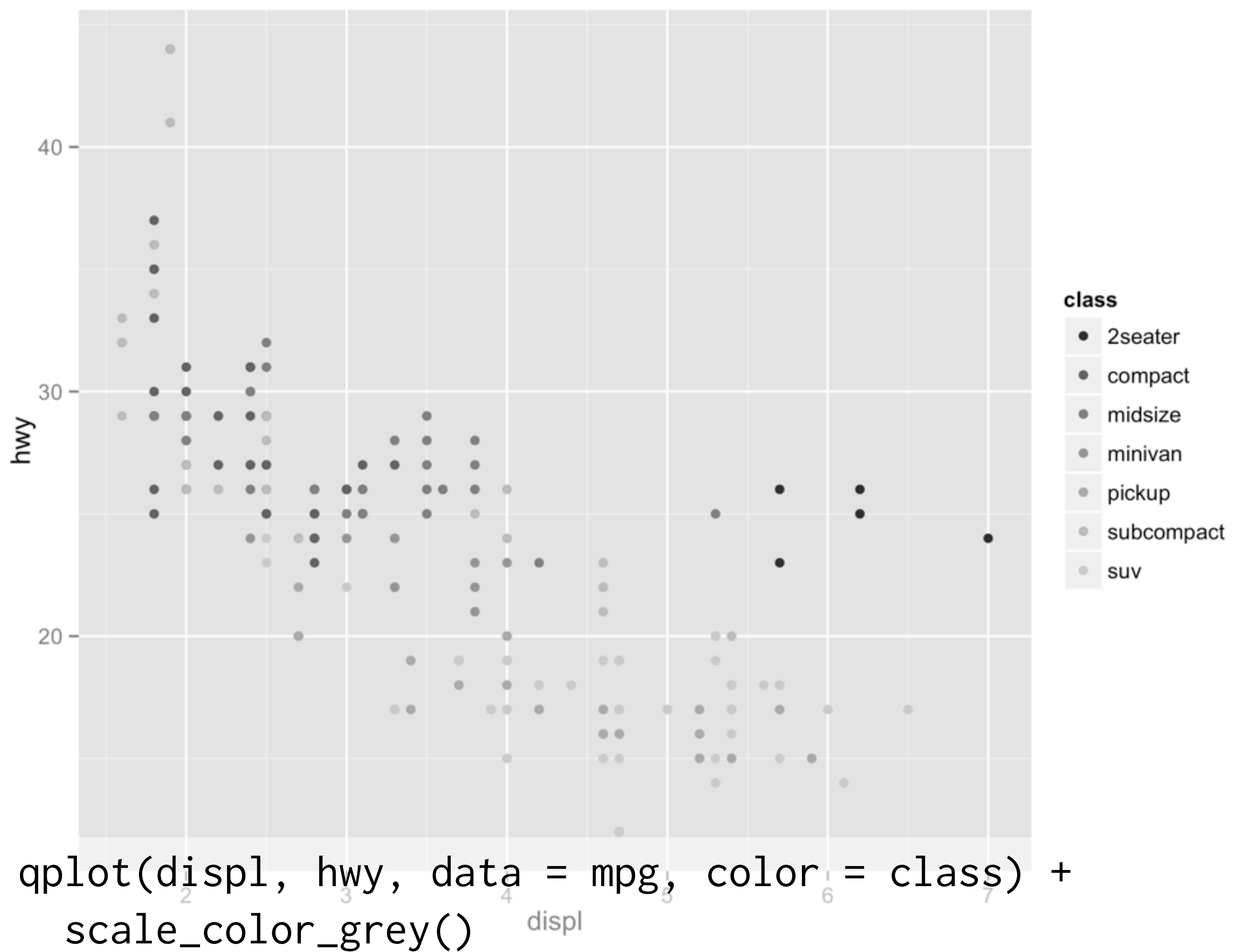
What variable to map to color

## **Scale**

How to map the variable to color



```
qplot(displ, hwy, data = mpg, color = class)
```



# Scales

The details of an aesthetic mapping.

Naming scheme: *scale\_aesthetic\_scalename*

```
qplot(displ, hwy, data = mpg, color = class) +  
  scale_color_grey()
```

# Scales

The details of an aesthetic mapping.

Naming scheme: *scale\_aesthetic\_scalename*

```
tx + scale_fill_grey()
```

# Scales

The details of an aesthetic mapping.

Naming scheme: `scale_aesthetic_scalename`

```
tx + scale_fill_grey()
```

# Scales

The details of an aesthetic mapping.

Naming scheme: `scale_aesthetic_scalename`

```
tx + scale_fill_grey()
```



# Scales

The details of an aesthetic mapping.

Naming scheme: `scale_aesthetic_`*scalename*

```
tx + scale_fill_grey()
```

# Defaults

By default, a scale is built for every aesthetic that your plot maps to data.

When you add a new scale, you override the default.

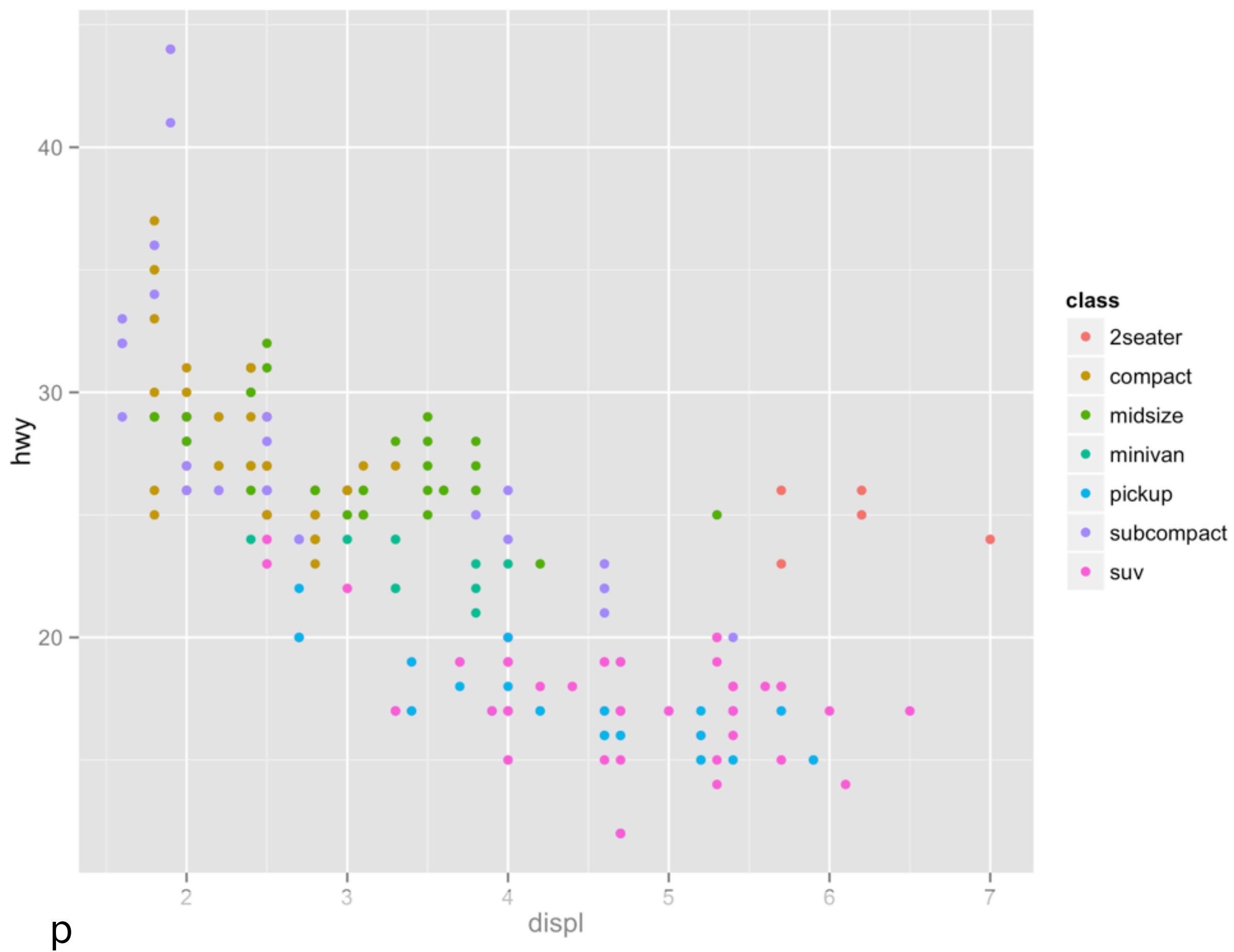
Most common scales to add:

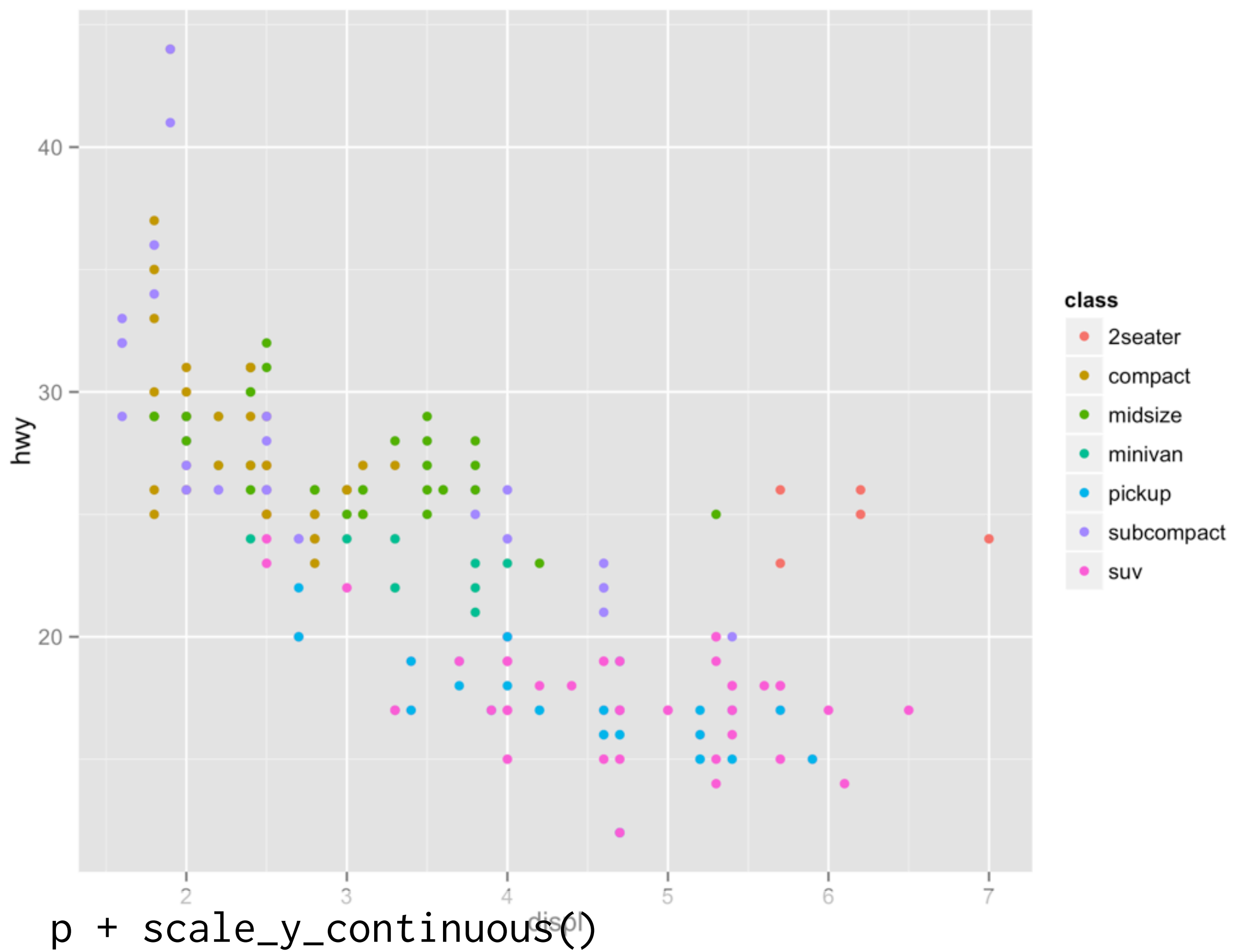
- `scale_aesthetic_`**continuous**, or
- `scale_aesthetic_`**discrete**

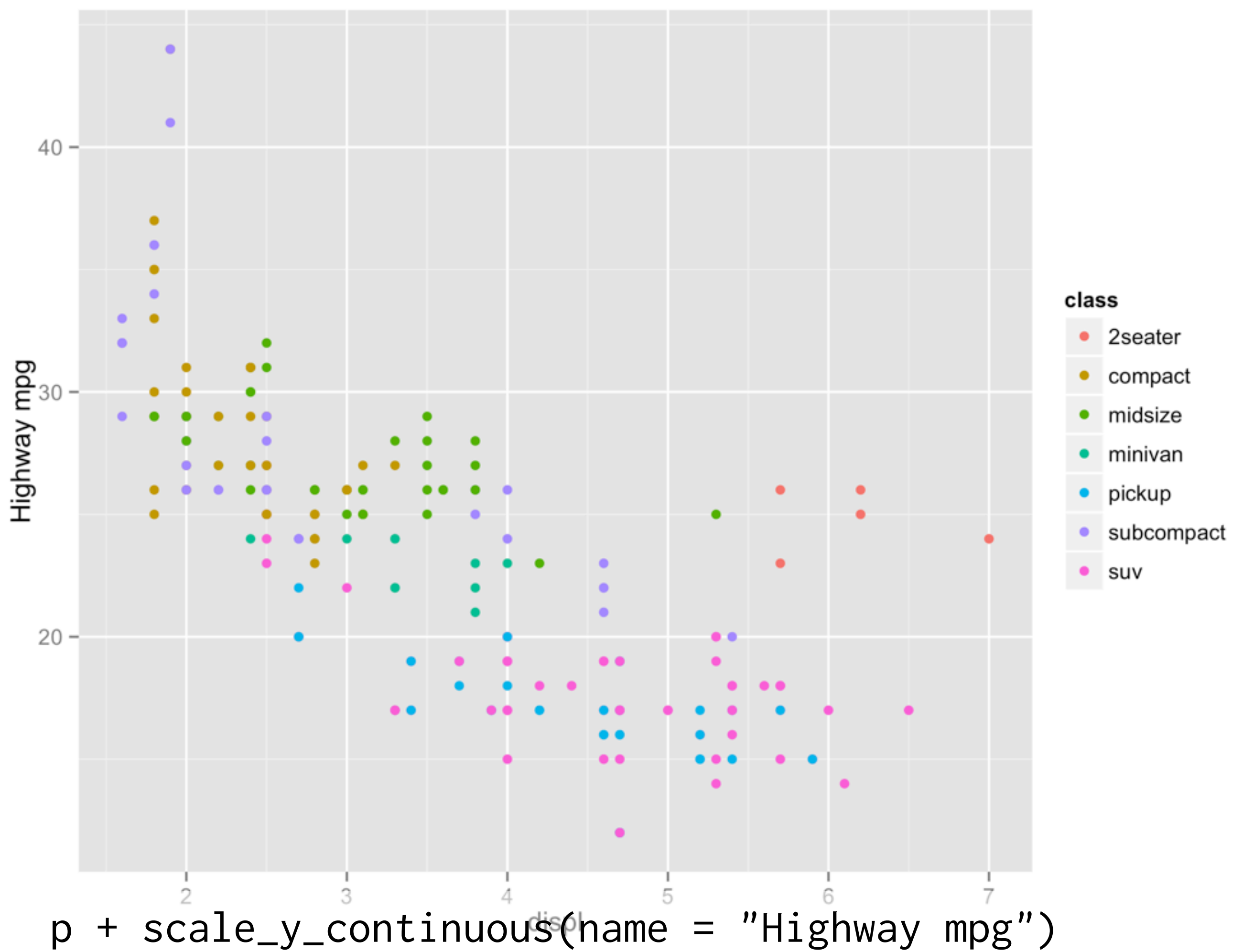
# What can you change with a scale?

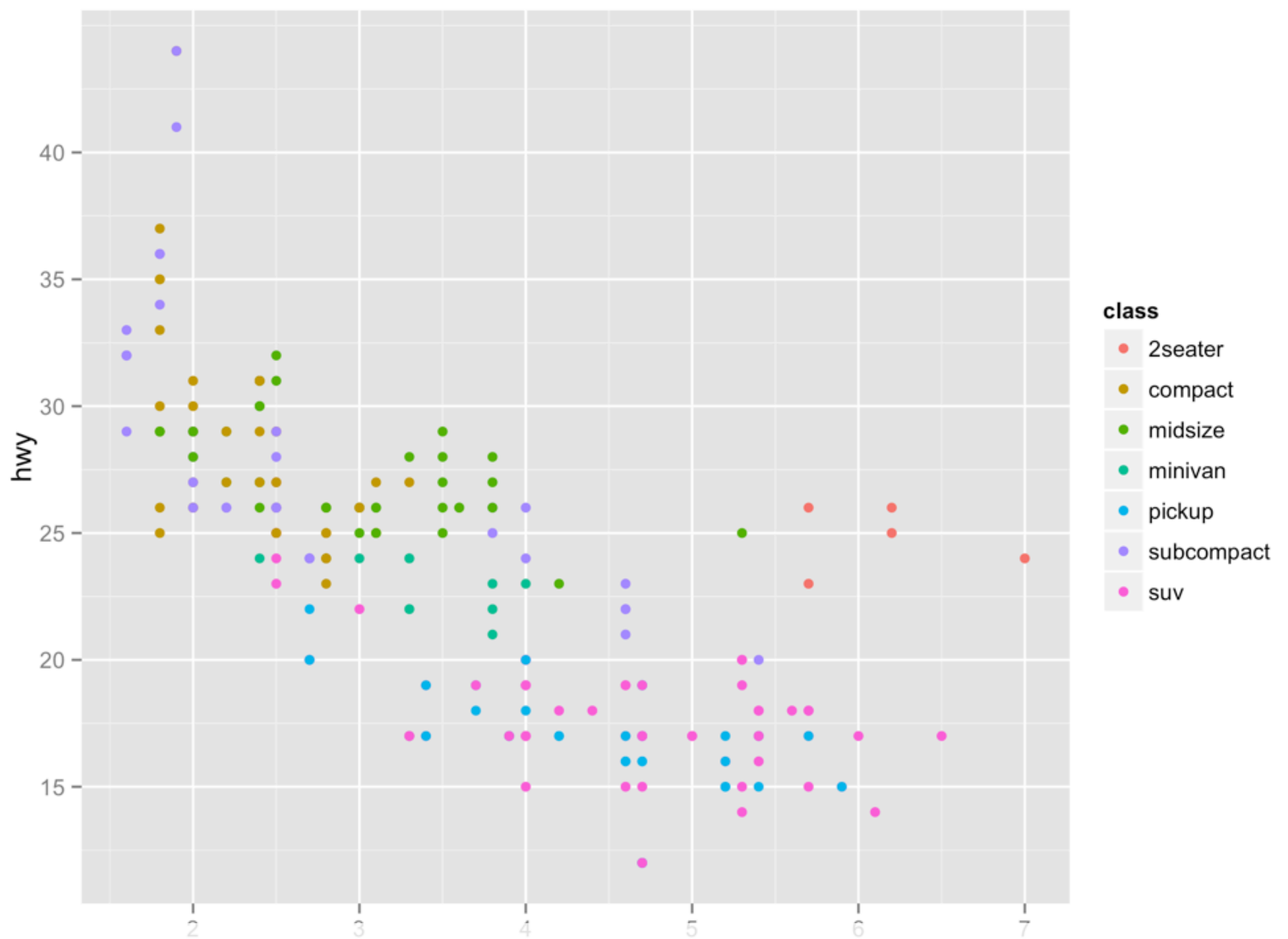
scale parameter	controls
name	axis labels (x, y) or legend title
breaks	where ticks occur (x, y) or legend entries
labels	tick labels (x, y) or legend labels
limits	The range of data to apply the mapping to

```
p <- qplot(displ, hwy, data = mpg, color = class)
```

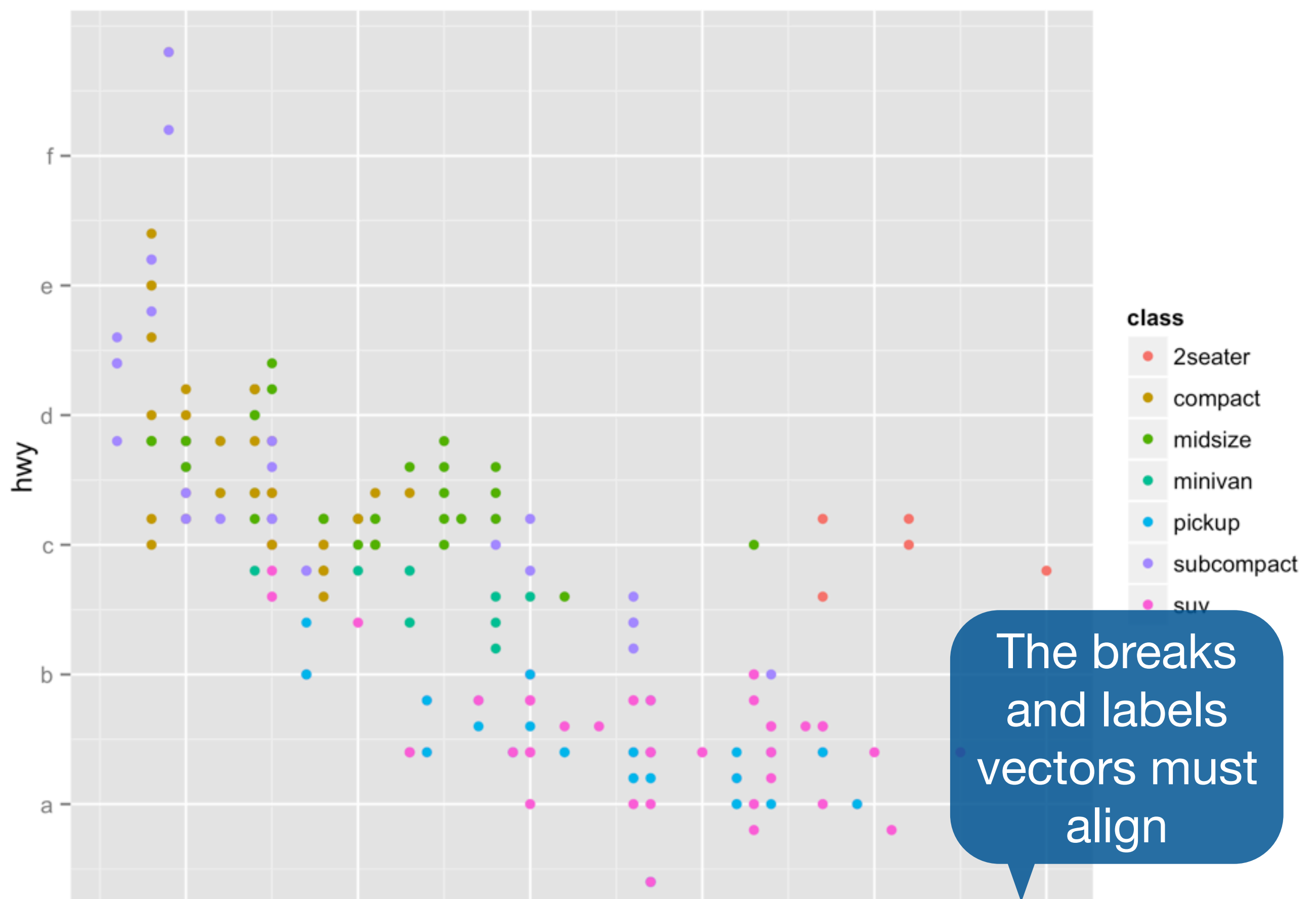






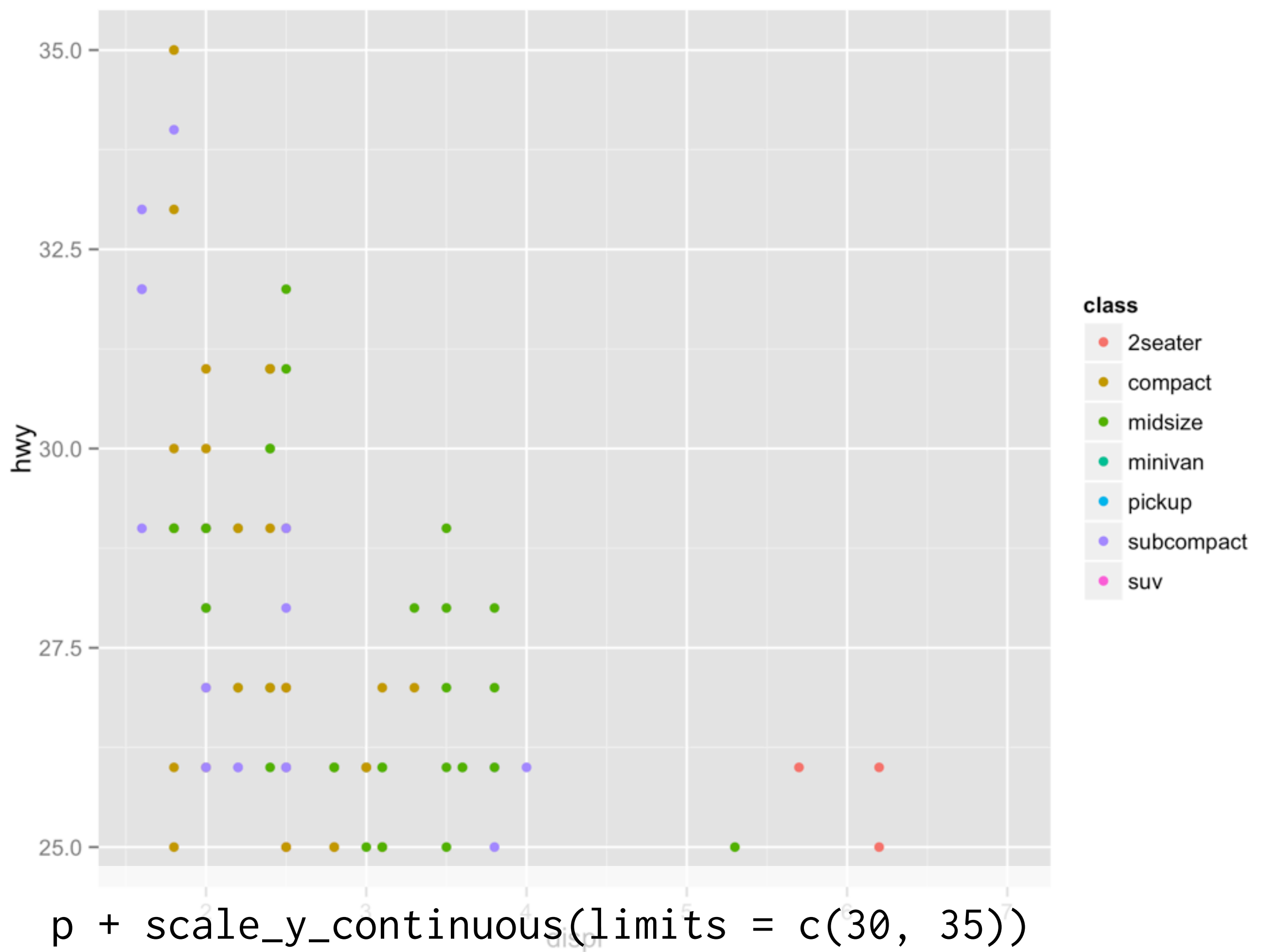


```
p + scale_y_continuous(breaks = c(15, 20, 25, 30, 35, 40))
```



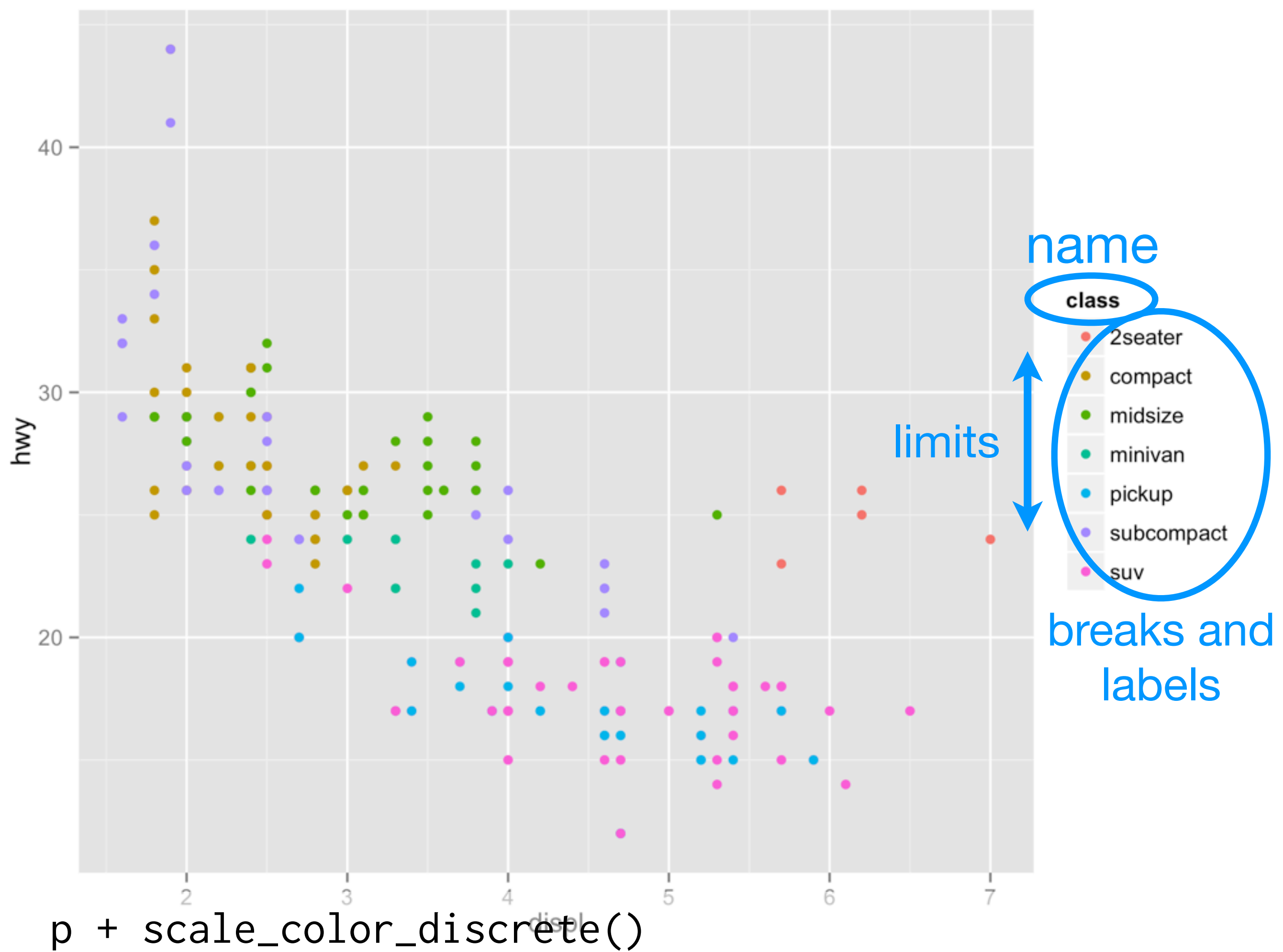
```
p + scale_y_continuous(breaks = c(15, 20, 25, 30, 35, 40),
  labels = c("a", "b", "c", "d", "e", "f"))
```





# Your turn

What happens if you add a scale for color (hint: `scale_color_discrete`) and change the name, breaks, labels, and limits parameters?



# Your turn

Use scales with `tx` to

1. remove the `long` and `lat` axis labels
2. change the title of the fill legend
3. create more informative legend labels

```
tx + scale_fill_discrete("Population",  
  labels = c("0 - 999", "1,000 - 9,999",  
    "10,000 - 99,999", "100,000 - 999,999",  
    "1,000,000+")) +  
scale_x_continuous("") +  
scale_y_continuous("")
```

# More exotic scales

A useful list of scales is available at <http://docs.ggplot2.org/current>

## Scales

Scales control the mapping between data and aesthetics.

- `expand_limits`  
Expand the plot limits with data.
- `guides`  
Set guides for each scale.
- `guide_legend`  
Legend guide.
- `guide_colourbar` (`guide_colorbar`)  
Continuous colour bar guide.
- `scale_alpha` (`scale_alpha_continuous`, `scale_alpha_discrete`)  
Alpha scales.
- `scale_area`  
Scale area instead of radius (for size).
- `scale_colour_brewer` (`scale_color_brewer`, `scale_fill_brewer`)  
Sequential, diverging and qualitative colour scales from colorbrewer.org
- `scale_colour_gradient` (`scale_color_continuous`, `scale_color_gradient`, `scale_colour_continuous`, `scale_fill_continuous`, `scale_fill_gradient`)  
Smooth gradient between two colours
- `scale_colour_gradient2` (`scale_color_gradient2`, `scale_fill_gradient2`)  
Diverging colour gradient
- `scale_colour_gradientn` (`scale_color_gradientn`, `scale_fill_gradientn`)  
Smooth colour gradient between n colours



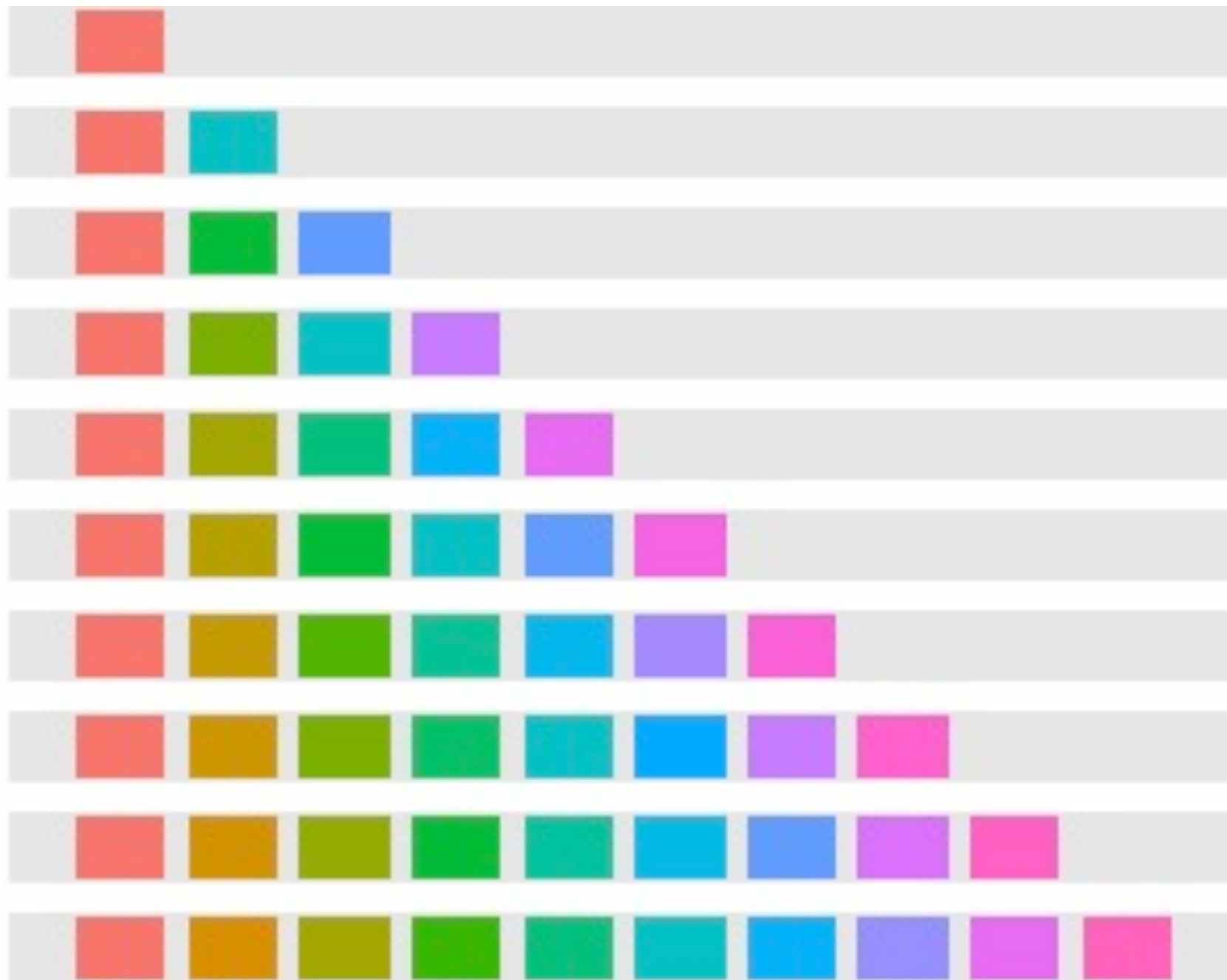
# Colors

**Colour** is the most popular aesthetic after position. It is also the easiest to misuse.

color **spaces**

color **blindness**.

# Default discrete palettes





# Default continuous palette



# Custom color scales

Discrete:

```
scale_fill_manual  
scale_fill_brewer  
scale_fill_grey
```

Continuous:

```
scale_fill_gradient  
scale_fill_gradient2  
scale_fill_gradientn
```

# *Manual* Discrete Scales

```
scale_color_manual  
scale_fill_manual  
scale_size_manual  
scale_shape_manual
```

new parameter	controls
values	values to use in the scale (specific colors, sizes, etc.)

```
r <- qplot(displ, cty, colour = drv, shape = fl,
  data = mpg)
```

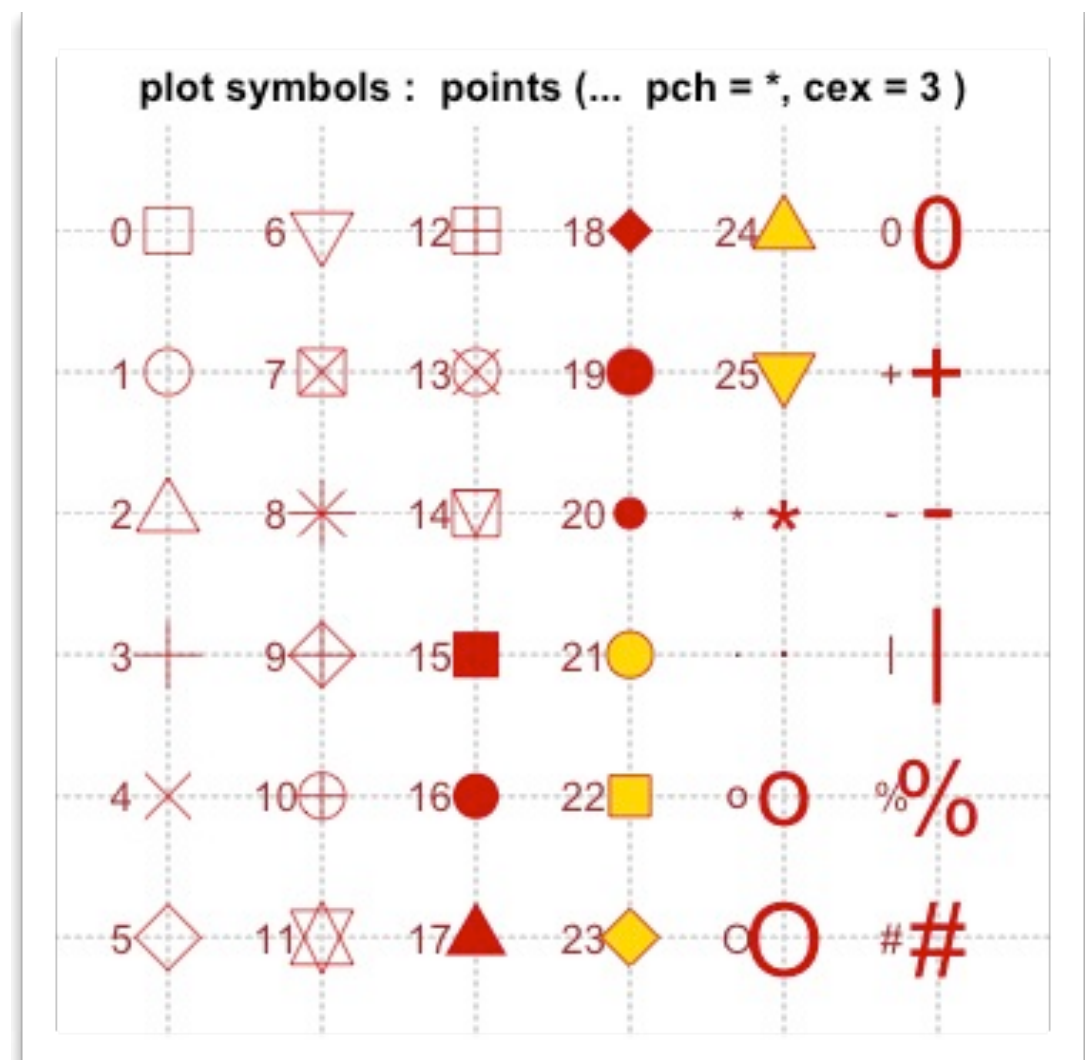
```
# Specify colors manually
```

```
r + scale_color_manual(values = c("red", "black",
  "#3333cc"))
```

```
# Specify the shapes manually
```

```
r + scale_shape_manual(
  values = c(0, 15, 1, 16, 3))
```

?pch



# Manual Scales

Offer complete control

Often look worse than you'd imagine

# *Brewer* Discrete Scales

```
scale_color_brewer  
scale_fill_brewer
```

new parameter	controls
<b>palette</b>	name of a palette in the RColorBrewer package

# Color brewer

Cynthia Brewer developed useful, pleasing palettes, particularly tailored for maps:

<http://colorbrewer2.org>

```
library(RColorBrewer)
```

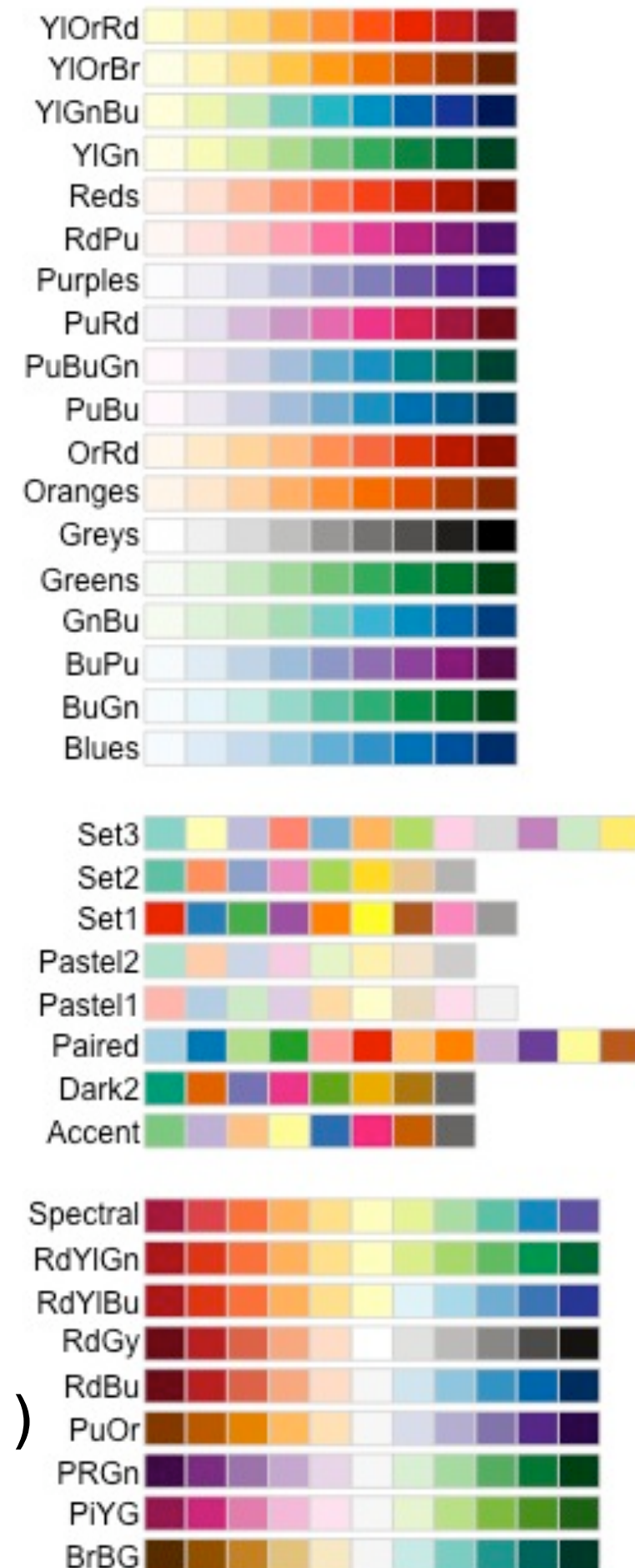
```
RColorBrewer::display.brewer.all()
```

```
q + scale_color_brewer(palette="Spectral")
```

```
q + scale_color_brewer(palette="Set3")
```

```
library(scales)
```

```
show_col(brewer_pal(palette = "YlOrRd")(9))
```



# Your turn

Modify the tx fill scale to use a palette that better maps population.

Use `RColorBrewer::display.brewer.all()` if you'd like to see possible brewer palletes.



```
tx + scale_fill_brewer("Population",  
  palette = "Blues",  
  labels = c("0 - 999", "1,000 - 9,999",  
    "10,000 - 99,999", "100,000 - 999,999",  
    "1,000,000+")) +  
scale_x_continuous("") +  
scale_y_continuous("")
```

# **Saving graphics**

# Your turn

What does this command return?

`getwd()`

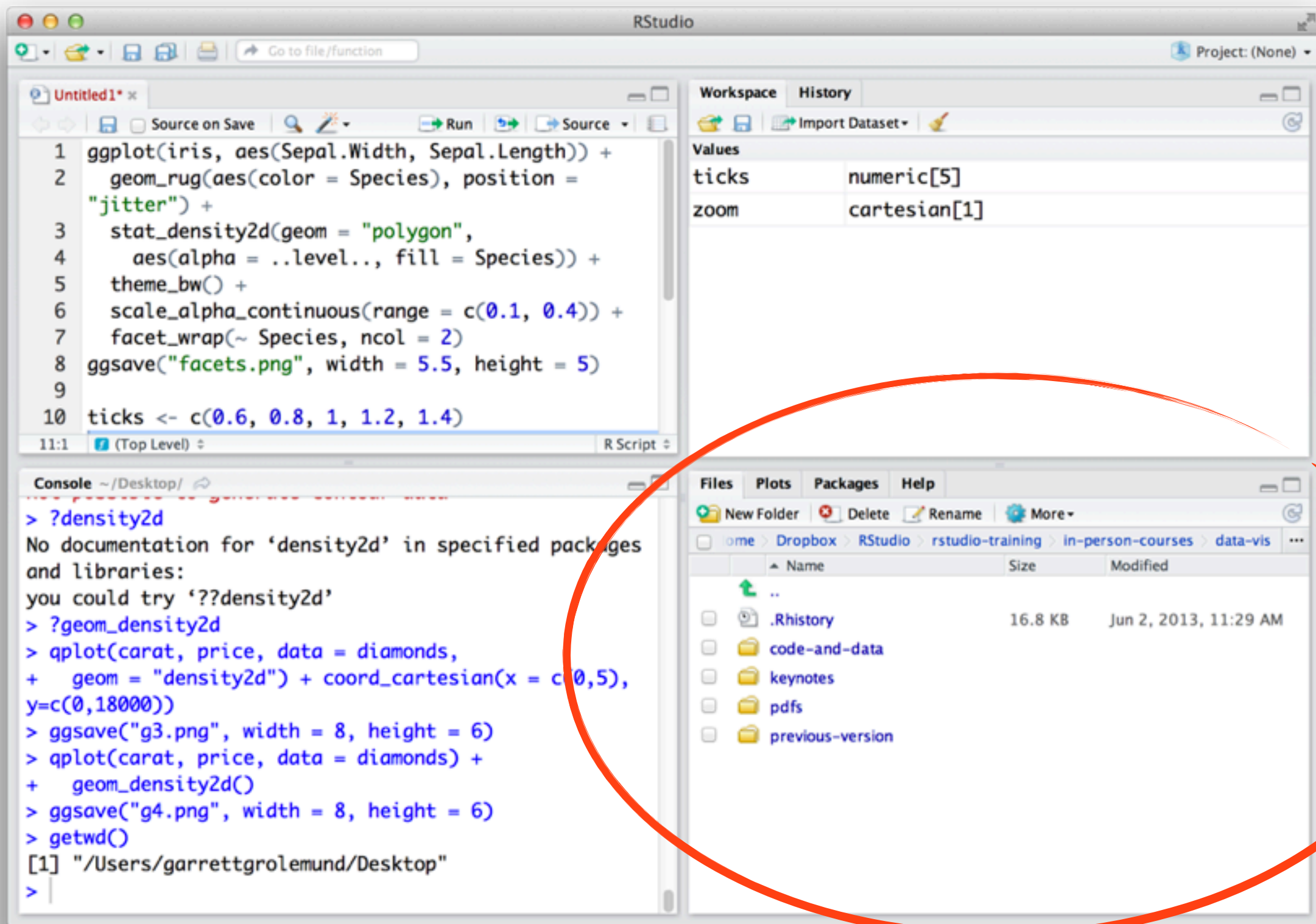
Could you find that file  
on your computer?

# Working directory

When you start R, it associates itself with a folder (i.e, directory) on your computer.

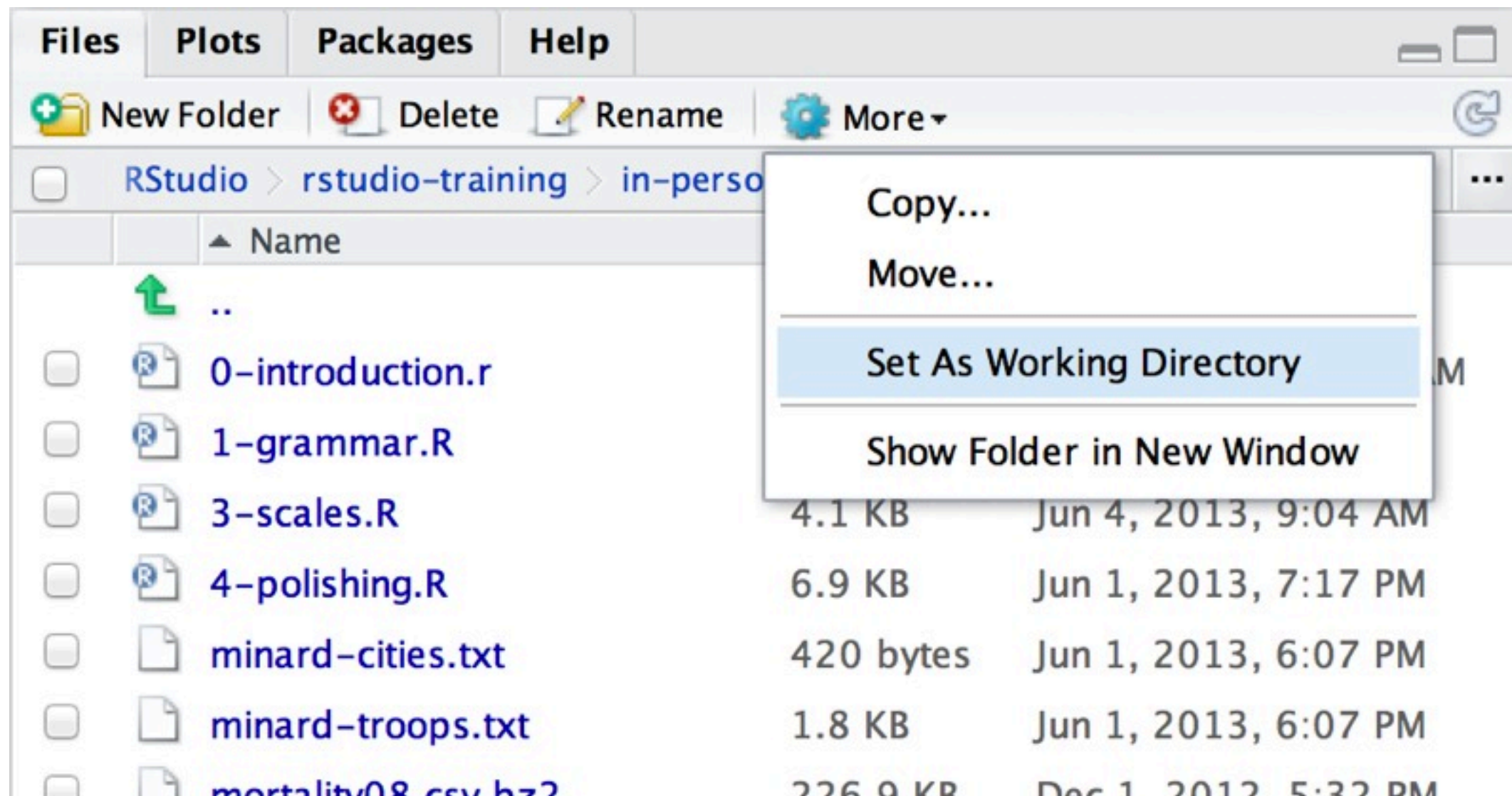
- This folder is known as your "**working directory**"
- When you save files, R will save them here
- When you load files, R will look for them here

The files pane of RStudio displays the contents of your working directory



# Changing the Working directory

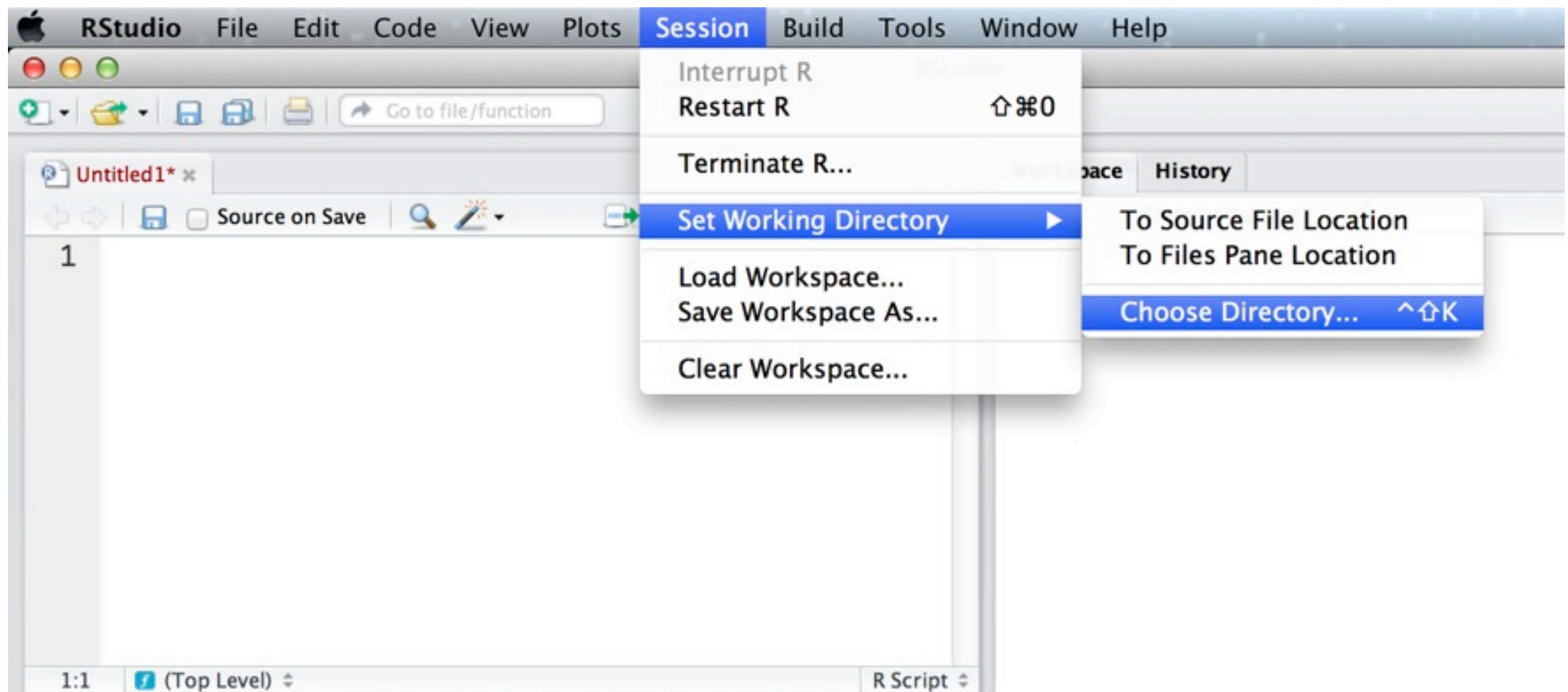
**First option:** Navigate in the files pane to a new directory. Click More>Set As Working Directory





# Changing the Working directory

**Second option:** In the toolbar, go to Session>Set Working Directory>Choose Directory...



# Your turn

Change your working directory to the folder you downloaded for today's course.

Note: this folder came as a .zip archive. You must extract the .zip file before you can use it as a directory.



# Saving plots

# Uses size on screen:

```
ggsave("my-plot.pdf")
```

```
ggsave("my-plot.png")
```

# Specify size in inches

```
ggsave("my-plot.pdf", width = 6, height = 6)
```

PDF	PNG
Vector based (can zoom in infinitely)	Raster based (made up of pixels)
Good for most plots	Good for plots with thousands of points

**Where to go  
from here**

## Help topics

### Geoms

Geoms, short for geometric objects, describe the type of plot you will produce.

- [geom\\_abline](#)  
Line specified by slope and intercept.
- [geom\\_area](#)  
Area plot.
- [geom\\_bar](#)  
Bars, rectangles with bases on x-axis
- [geom\\_bin2d](#)  
Add heatmap of 2d bin counts.
- [geom\\_blank](#)  
Blank, draws nothing.
- [geom\\_boxplot](#)  
Box and whiskers plot.
- [geom\\_contour](#)  
Display contours of a 3d surface in 2d.
- [geom\\_crossbar](#)  
Hollow bar with middle indicated by horizontal line.
- [geom\\_density](#)  
Display a smooth density estimate.
- [geom\\_density2d](#)  
Contours from a 2d density estimate.
- [geom\\_dotplot](#)  
Dot plot



## Dependencies

- **Depends:** stats, methods
- **Imports:** plyr, digest, grid, gtable, reshape2, scales, memoise, proto, MASS
- **Suggests:** quantreg, Hmisc, mapproj, maps, hexbin, maptools, multcomp, nlme, testthat
- **Extends:** sp

# Learning ggplot2

**ggplot2 mailing list**

<http://groups.google.com/group/ggplot2>

**stackoverflow**

<http://stackoverflow.com/tags/ggplot2>

**Cookbook for common graphics**

<http://wiki.stdout.org/rcookbook/Graphs/>

**ggplot2 book**

<http://amzn.com/0387981403>

