

The art of flipbooking (building code movies)

With flipbookr and xaringan

Gina Reynolds, December 2019

Welcome

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- Yes, there is now a package called `flipbookr`
- Yes, it's still under construction (the title of this book used to be the "fragile perilous art")
- Yes, you can help make it less fragile and perilous by letting us know if/how it breaks and frustrates you and/or contributing at <https://github.com/EvaMaeRey/flipbookr!>

Spread the word and giving feedback

Please help us spread the word about flipbooks. Let your audience know *how* you created your flipbook with a quick acknowledgement, for example, *The flipbooked portion of this presentation was created with the new {flipbookr} package. Get it at remotes::install_github("EvaMaeRey/flipbookr")*

Also consider sharing your work on social media, and let me know what you've built on Twitter with a mention to @EvaMaeRey

Feedback? Contributions? Leave an issue at: <https://github.com/EvaMaeRey/flipbookr>

"Flipbooks"?

“Flipbooks” are tools that present side-by-side, aligned, incremental code-output evolution via automated code parsing and reconstruction. More about Flipbooks [here](#). There now exists a package for making Flipbooks for R: `flipbookr`. This is under development, but you are welcome to try it out by installing from github:

```
devtools::install_github("EvaMaeRey/flipbookr")
```

You can see the template that was used to build this flipbook that you are looking at right now [here](#).

Or, once you install the package (and restart RStudio?) a template for making the flipbook that you are looking at will also be available from within RStudio, File -> New File -> R Markdown -> From Template -> "A Minimal Flipbook".

How Flipbooking with Xaringan works

The flipbook you will be building here uses a member of the `rmarkdown` family called `Xaringan (presentation ninja)`, which creates a slideshow in html. Dynamic documents like `rmarkdown` documents allow you to comingle code and prose in a single document.

It may be obvious by now, if you are following along with the source template, that slide breaks are indicated with `---` (be careful trailing white space is not allowed)

Flipbooks are built by spawning new *partial* code chunks from a single, user-input code chunk. The partial code chunks build up and are display consecutively in a slide show alongside its output which yields a movie-like experience; this should make each step easier to understand.

As you begin with flipbooks, I'd recommend using the code chunk option `include = F` for your "source" code chunks, and with no caching throughout. As you begin to get more comfortable with flipbooking, you might change these choices.

Set-up

We use the flipbookr package, of course! This does the work of disassembling a single code chunk and creating the "build" of multiple partial-code chunks. This is at the top of this file in the "setup" code chunk.

Also, at the top of this template in that "setup" code chunk, I set *code chunk* options for the code chunks that follow. These will apply to the spawned code chunks.

```
Error in file(filename, "r", encoding = encoding) :  
  cannot open the connection
```

```
Error in file(filename, "r", encoding = encoding) :  
  cannot open the connection
```

Using `flipbookr::chunk_reveal()`

You will use the `chunk_reveal()` function **inline** to generate the derivative code chunks, rather than inside of a code chunk, so that the text that is generated is interpreted correctly when rendered. The inline code will look something like this:

```
`r chunk_reveal(chunk_name = "cars", break_type = "user")`
```

There are several modalities that you might be interested in using for "flipbookifying" your code and the next section is dedicated to demoing some of them below.

- **break type** -- *which lines of code should be revealed when*, `break_type` defaults to "auto"
- **display type** -- *display code and output, or just output, or just code?*, `display_type` defaults to "both"
- **assignment type** -- *does code chunk use left assignment?*, `left_assign` defaults to FALSE

At first we'll apply our flipbooking to the below input code - the code chunk is named "cars". For now I set `echo = TRUE` for this code chunk, so you can see the code content but sometimes you might like to set `echo` to `FALSE`. This code uses tidyverse tools, so we loaded that too in the "setup" code chunk at the beginning of the template.

break_type

Notice the regular comments and the special `#BREAK` comments, these will be used for a couple of the different "break type" modalities.

```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist) + #BREAK  
  geom_point(  
    alpha = .3, #BREAK2  
    color = "blue" #BREAK3  
  ) + #BREAK  
  aes(size = speed) #BREAK
```

break_type = "auto"

One parameter of flipbooking is the `break_type`. The default is "auto", in which appropriate breakpoints are determined automatically --- by finding where parentheses are balanced.

```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist) + #BREAK  
  geom_point(  
    alpha = .3, #BREAK2  
    color = "blue" #BREAK3  
  ) + #BREAK  
  aes(size = speed) #BREAK
```

cars

speed dist

1	4	2
2	4	10
3	7	4
4	7	22
5	8	16
6	9	10
7	10	18
8	10	26
9	10	34
10	11	17
11	11	28
12	12	14
13	12	20
14	12	24
15	12	28
16	13	26
17	13	34
18	13	34
19	13	46
20	14	26
21	14	36
22	14	60
23	14	80
24	15	20
25	15	26
26	15	54
27	16	32
28	16	40
29	17	32
30	17	40
31	17	50
32	18	42

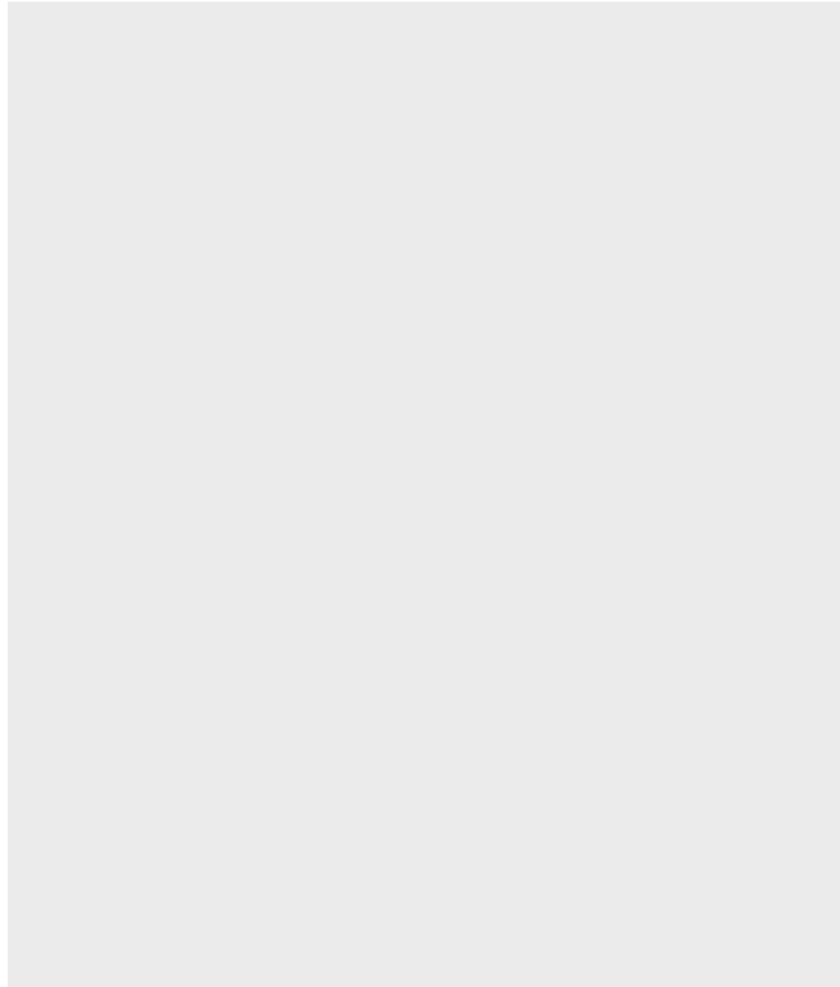
```
cars %>%
```

```
  filter(speed > 4)
```

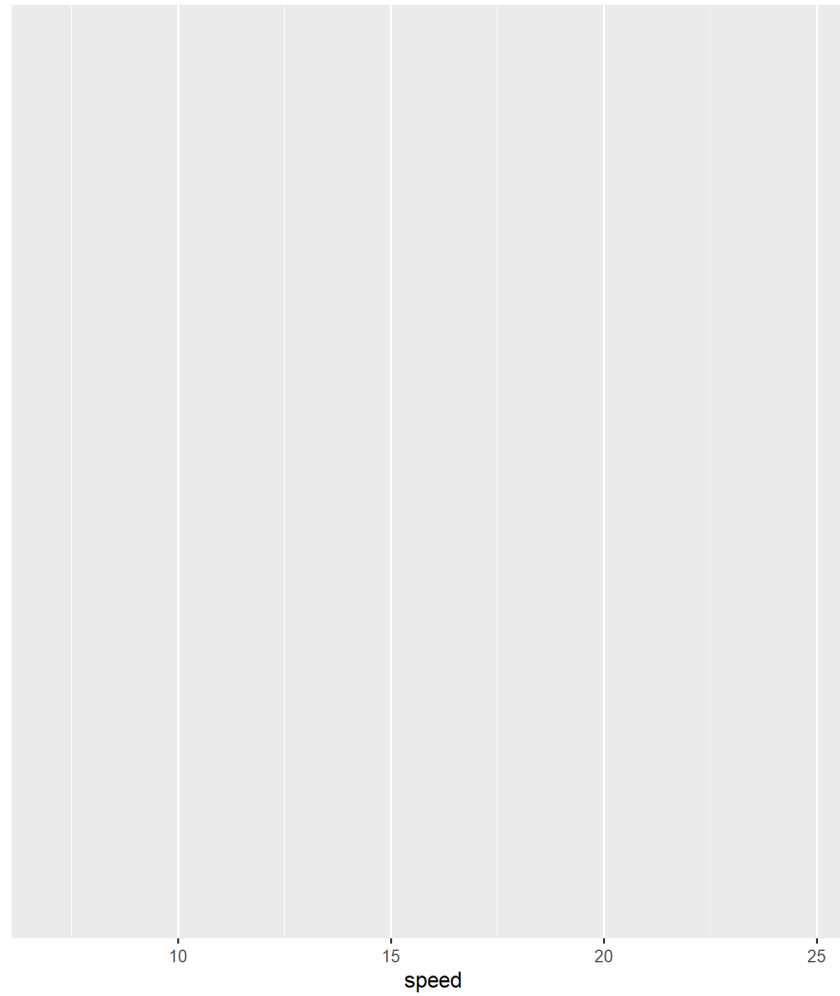
```
speed dist
```

1	7	4
2	7	22
3	8	16
4	9	10
5	10	18
6	10	26
7	10	34
8	11	17
9	11	28
10	12	14
11	12	20
12	12	24
13	12	28
14	13	26
15	13	34
16	13	34
17	13	46
18	14	26
19	14	36
20	14	60
21	14	80
22	15	20
23	15	26
24	15	54
25	16	32
26	16	40
27	17	32
28	17	40
29	17	50
30	18	42
31	18	56
32	18	76

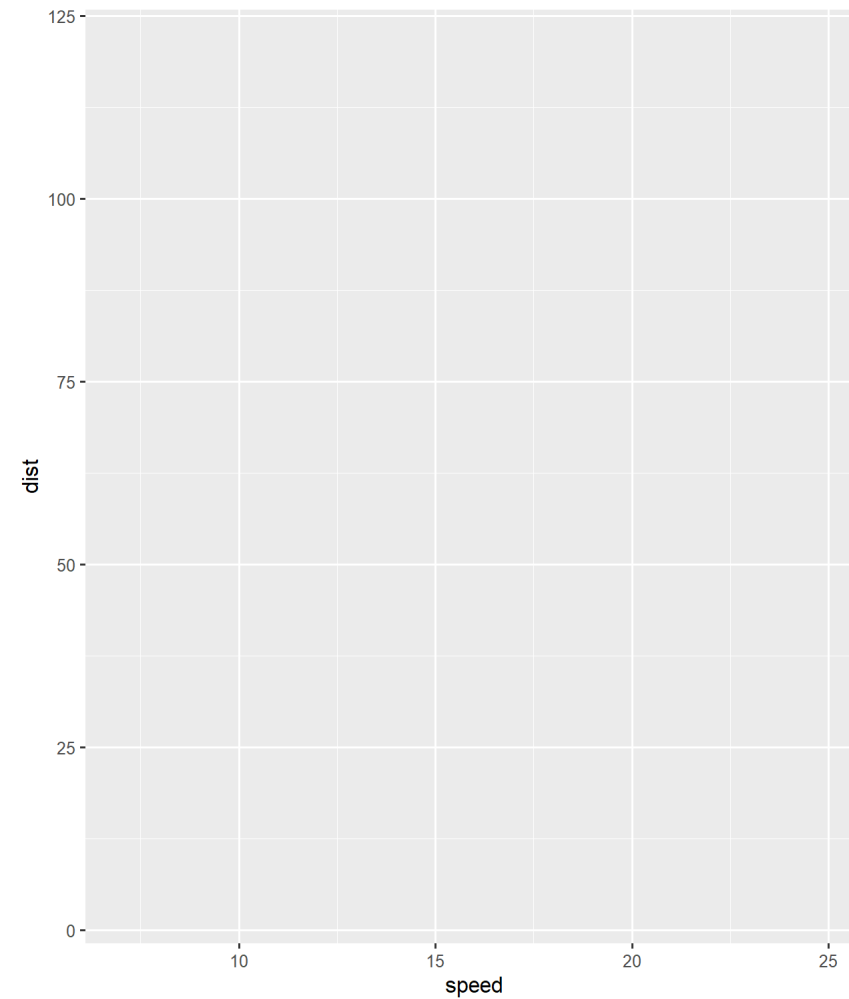
```
cars %>%  
  filter(speed > 4) %>%  
  ggplot()
```



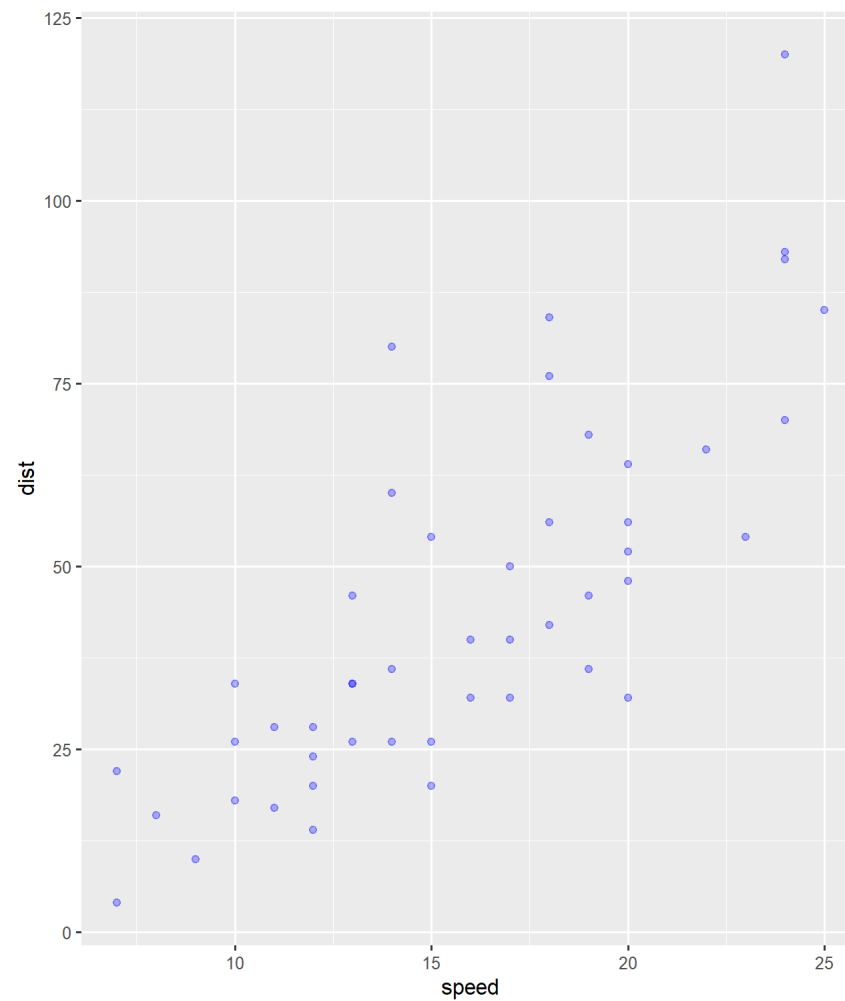

```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed)
```



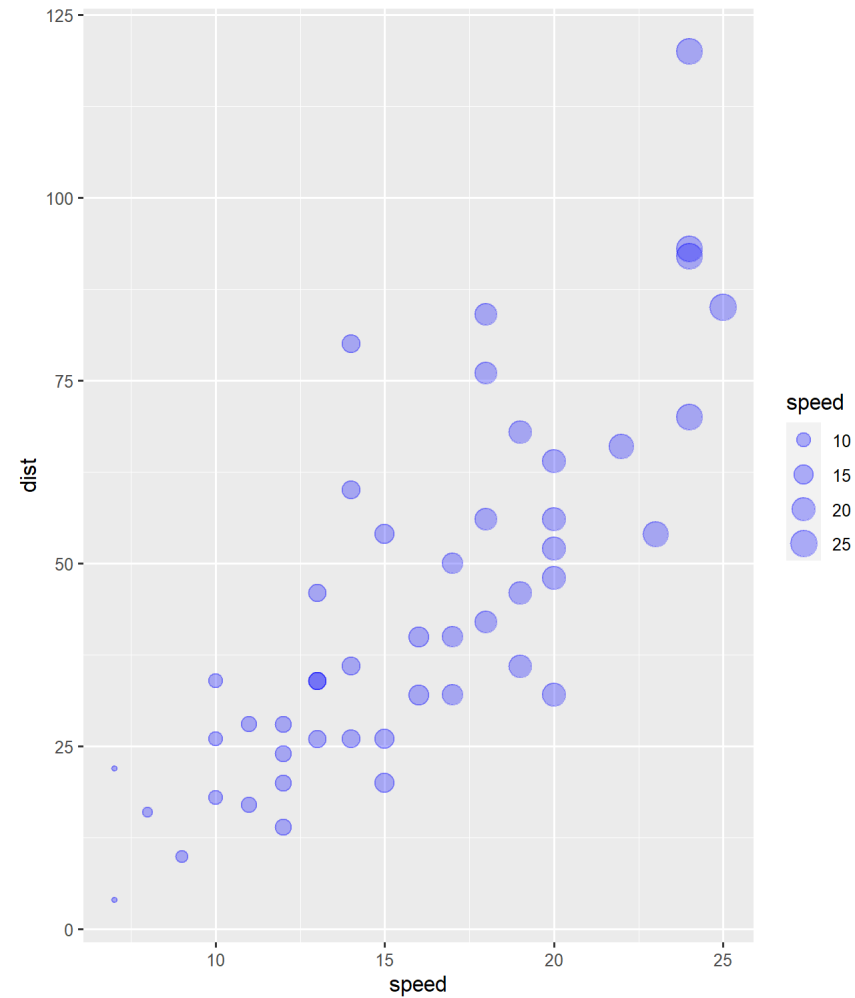
```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist)
```



```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist) +  
  geom_point(  
    alpha = .3,  
    color = "blue"  
  )
```



```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist) +  
  geom_point(  
    alpha = .3,  
    color = "blue"  
  ) +  
  aes(size = speed)
```

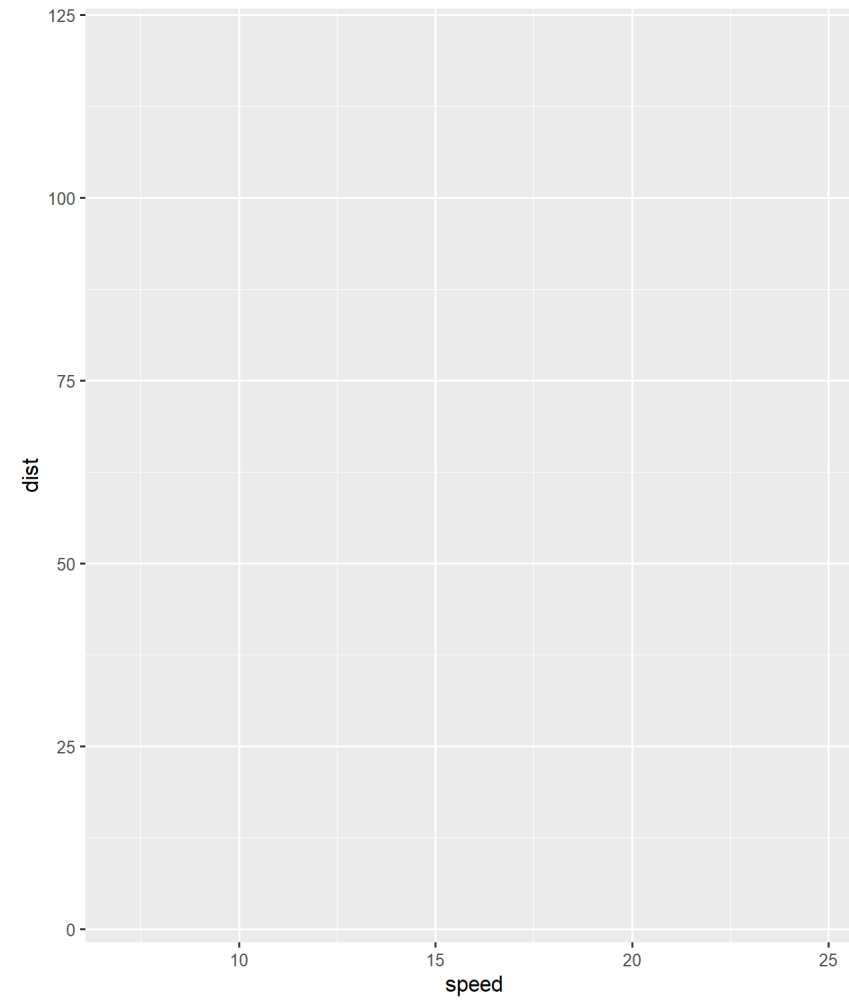


break_type = "user", with #BREAK

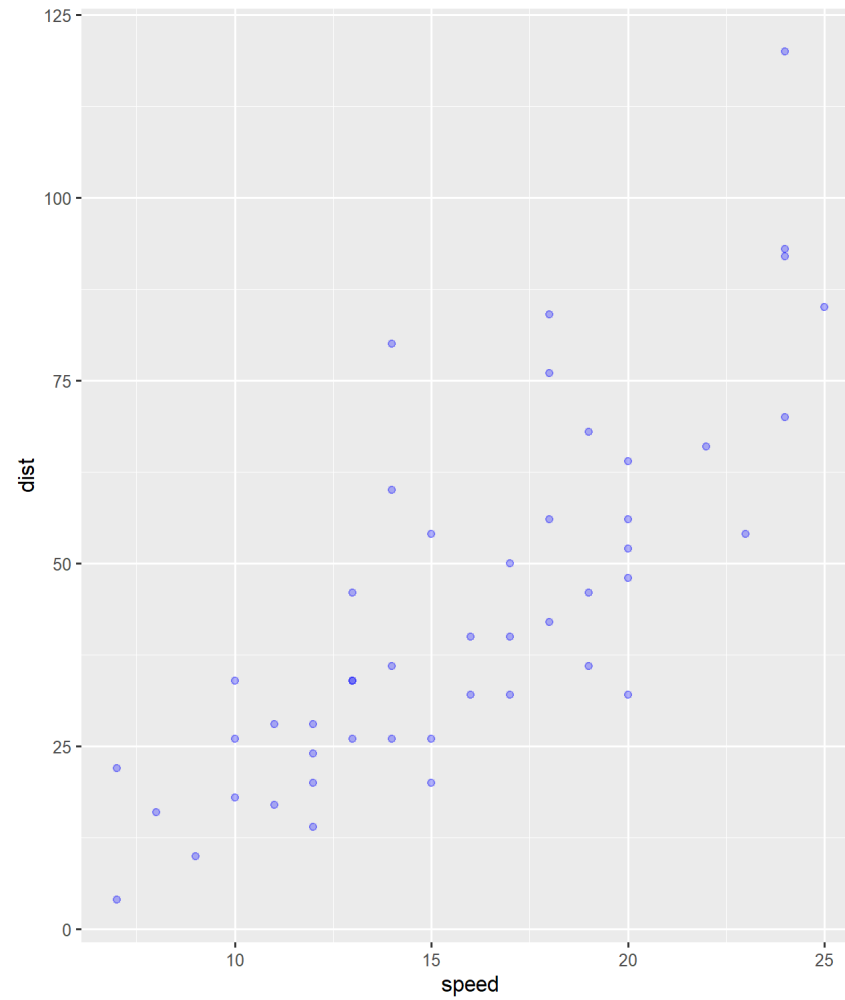
If the break_type is set to "user", the breakpoints are those indicated by the user with the special comment #BREAK

```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist) + #BREAK  
  geom_point(  
    alpha = .3, #BREAK2  
    color = "blue" #BREAK3  
  ) + #BREAK  
  aes(size = speed) #BREAK
```

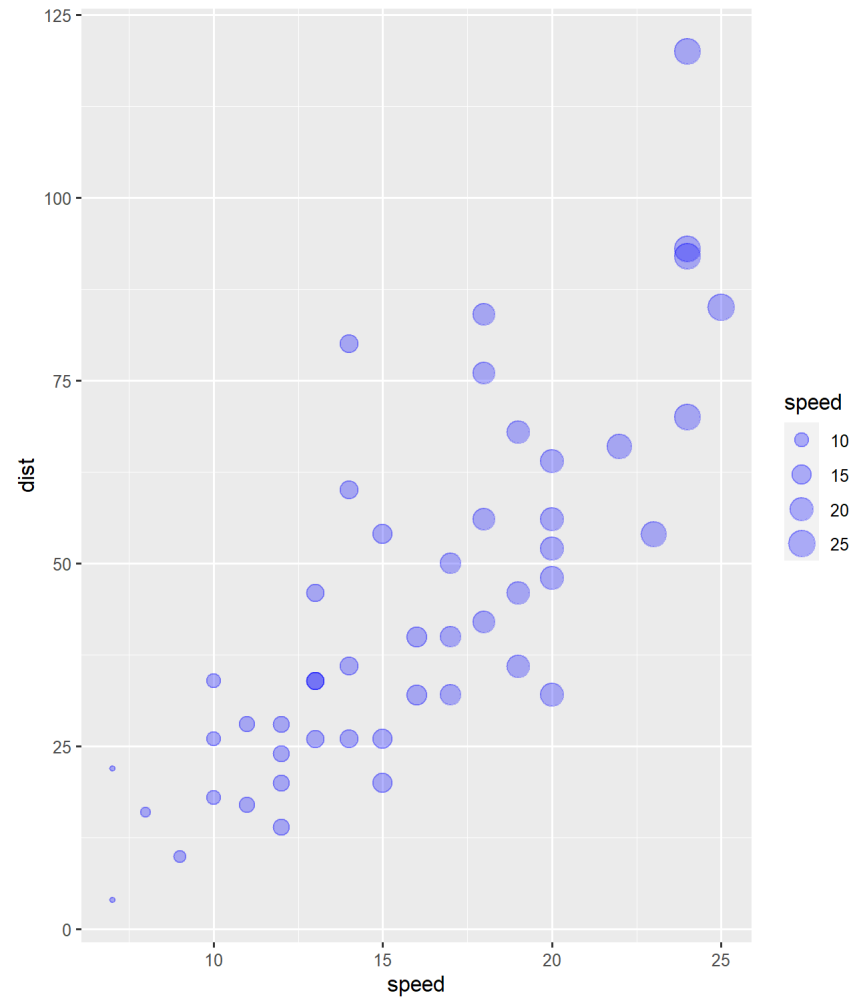
```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist)
```



```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist) +  
  geom_point(  
    alpha = .3,  
    color = "blue"  
  )
```



```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist) +  
  geom_point(  
    alpha = .3,  
    color = "blue"  
  ) +  
  aes(size = speed)
```

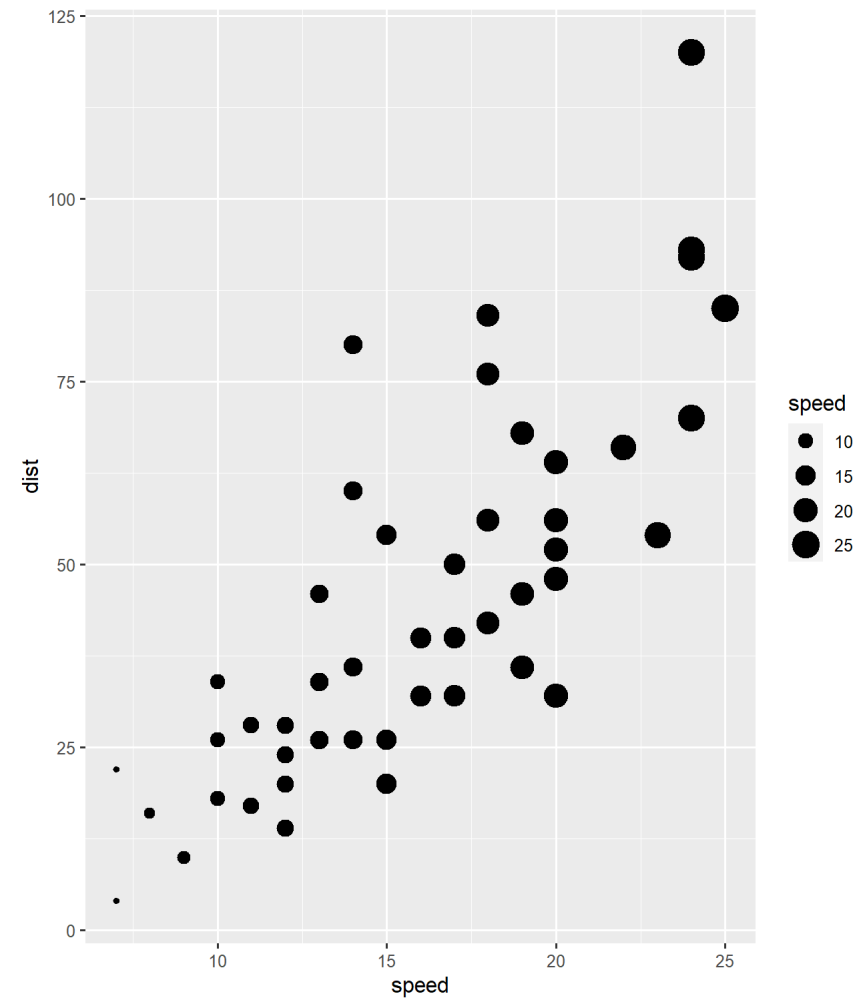


break_type = "non_seq", with #BREAK2, #BREAK3

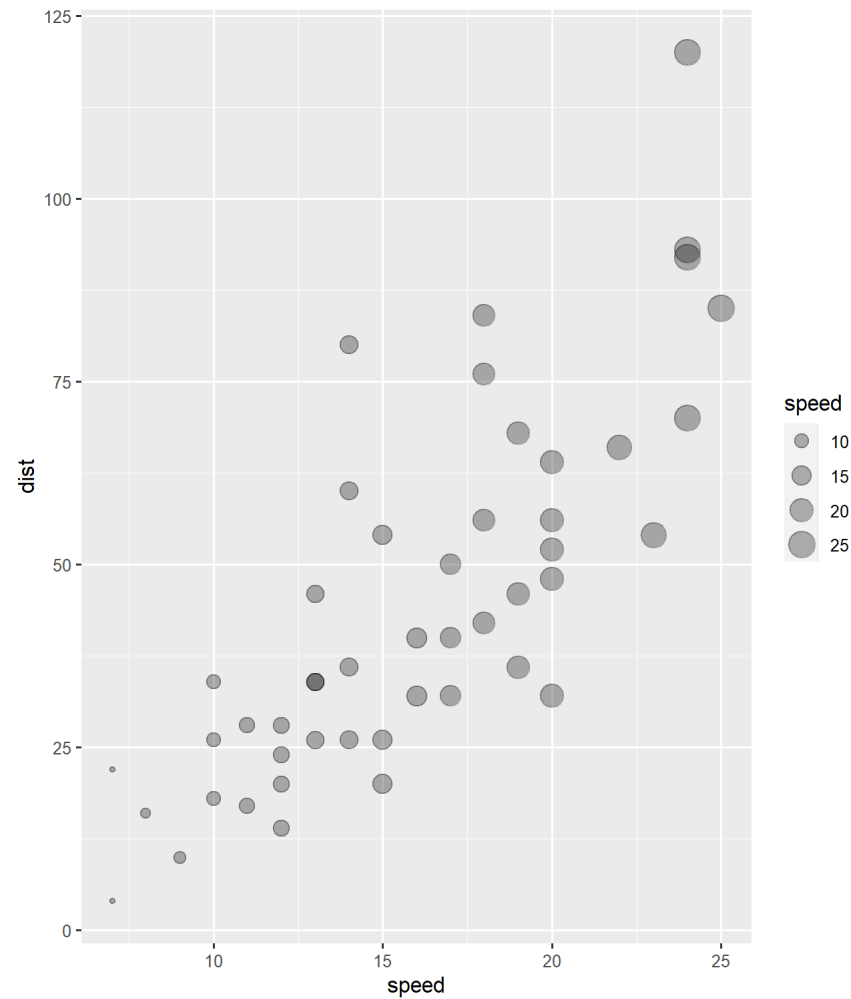
If the break_type is set to "non_seq", the breakpoints are those indicated by the user with the special numeric comment #BREAK2, #BREAK3 etc to indicate at which point in time the code should appear.

```
cars %>%
  filter(speed > 4) %>%
  ggplot() +
  aes(x = speed) +
  aes(y = dist) + #BREAK
  geom_point(
    alpha = .3, #BREAK2
    color = "blue" #BREAK3
  ) + #BREAK
  aes(size = speed) #BREAK
```

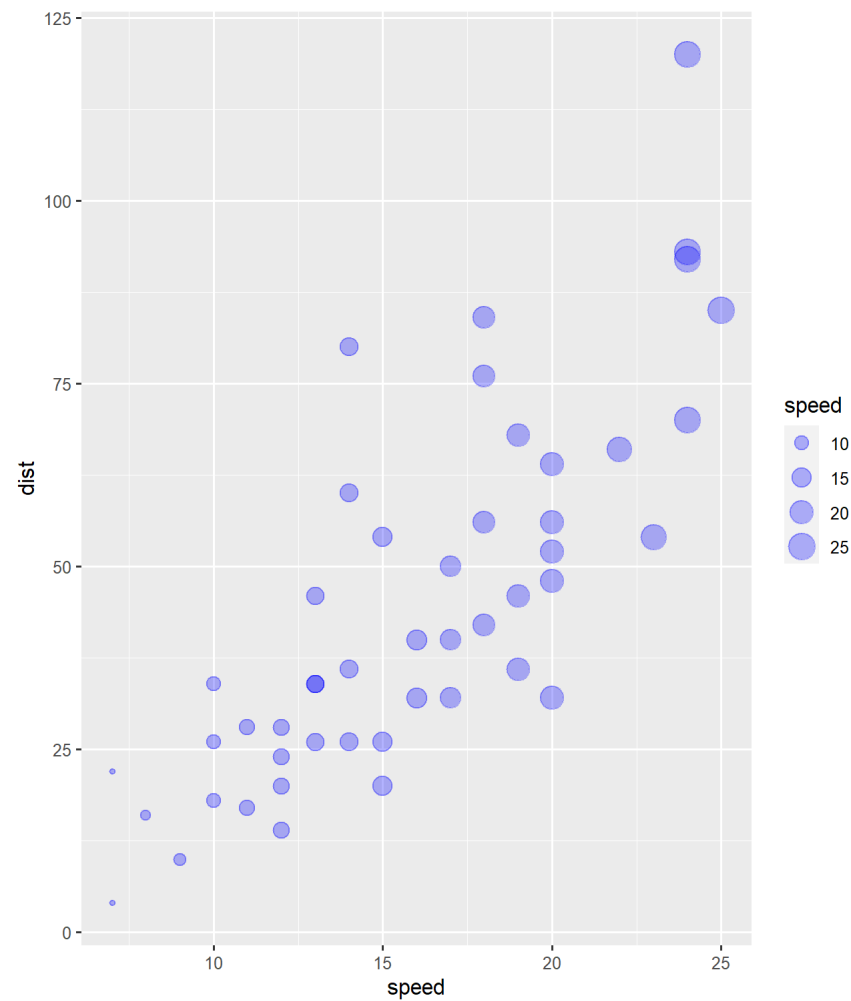
```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist) +  
  geom_point(  
    ) +  
  aes(size = speed)
```



```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist) +  
  geom_point(  
    alpha = .3,  
  ) +  
  aes(size = speed)
```



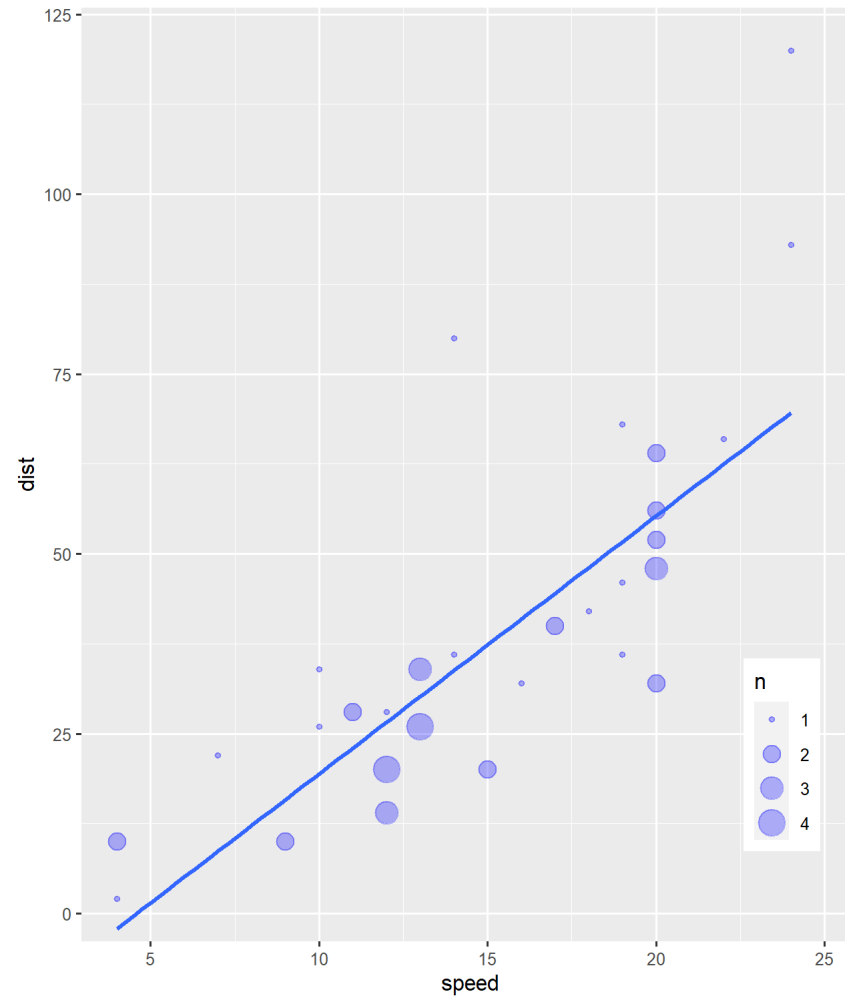
```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist) +  
  geom_point(  
    alpha = .3,  
    color = "blue"  
  ) +  
  aes(size = speed)
```



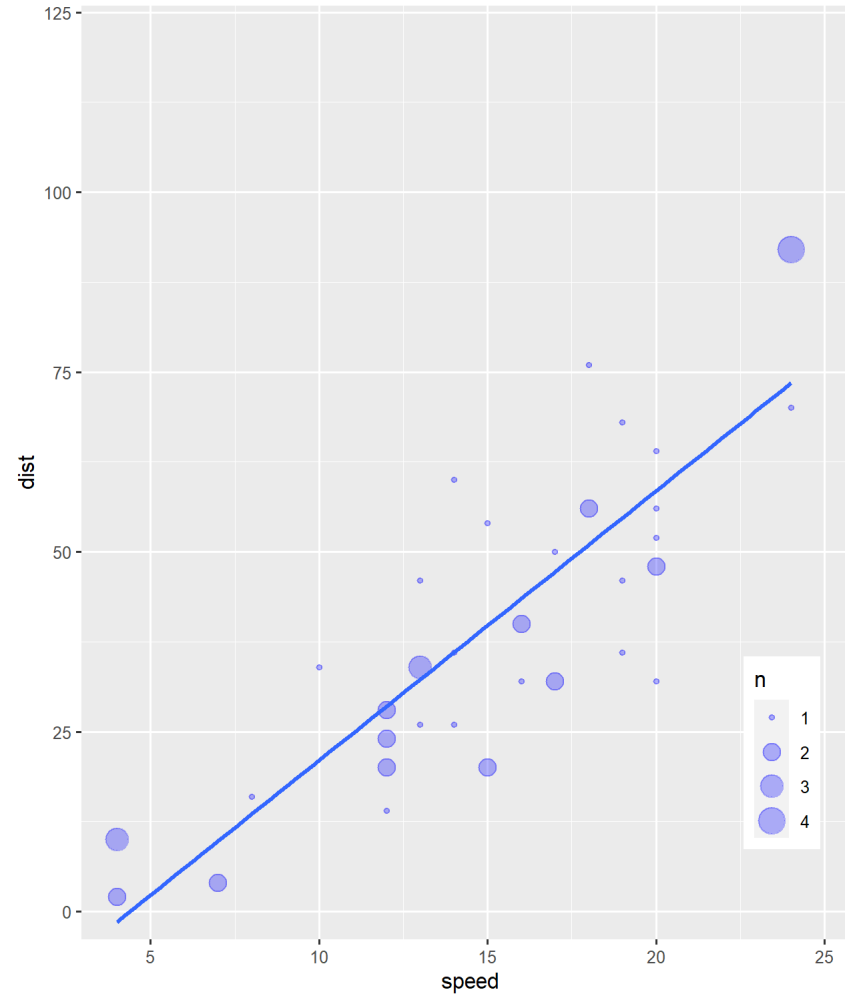
break_type = 5 (or "entering the multiverse")

Another modality is to set `break_type` equal to a positive integer, indicating that you want the same code chunk to be displayed multiple times. This makes the most sense in a setting where there is some randomization or random sampling and you want to see different realizations. Let's see this used on the user input code chunk "cars_multi", whose first step is to randomly sample rows from the data set cars with replacement.

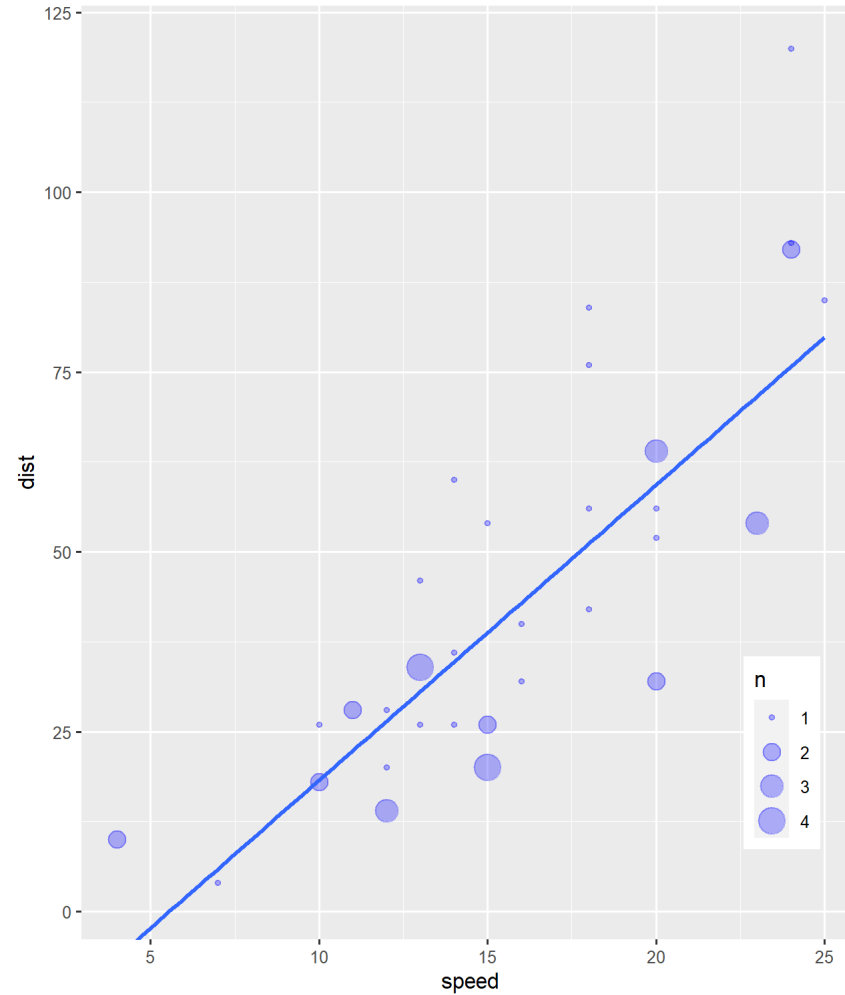
```
cars %>%
  sample_frac(size = 1, replace = TRUE) %>%
  ggplot() +
  aes(x = speed) +
  aes(y = dist) +
  geom_count(
    alpha = .3,
    color = "blue"
  ) +
  geom_smooth(method = lm, se = FALSE) +
  coord_cartesian(xlim = range(cars$speed),
                  ylim = range(cars$dist)) +
  theme(legend.position = c(.9, .2))
```



```
cars %>%
  sample_frac(size = 1, replace = TRUE) %>%
  ggplot() +
  aes(x = speed) +
  aes(y = dist) +
  geom_count(
    alpha = .3,
    color = "blue"
  ) +
  geom_smooth(method = lm, se = FALSE) +
  coord_cartesian(xlim = range(cars$speed),
                  ylim = range(cars$dist)) +
  theme(legend.position = c(.9, .2))
```



```
cars %>%
  sample_frac(size = 1, replace = TRUE) %>%
  ggplot() +
  aes(x = speed) +
  aes(y = dist) +
  geom_count(
    alpha = .3,
    color = "blue"
  ) +
  geom_smooth(method = lm, se = FALSE) +
  coord_cartesian(xlim = range(cars$speed),
                  ylim = range(cars$dist)) +
  theme(legend.position = c(.9, .2))
```



display_type

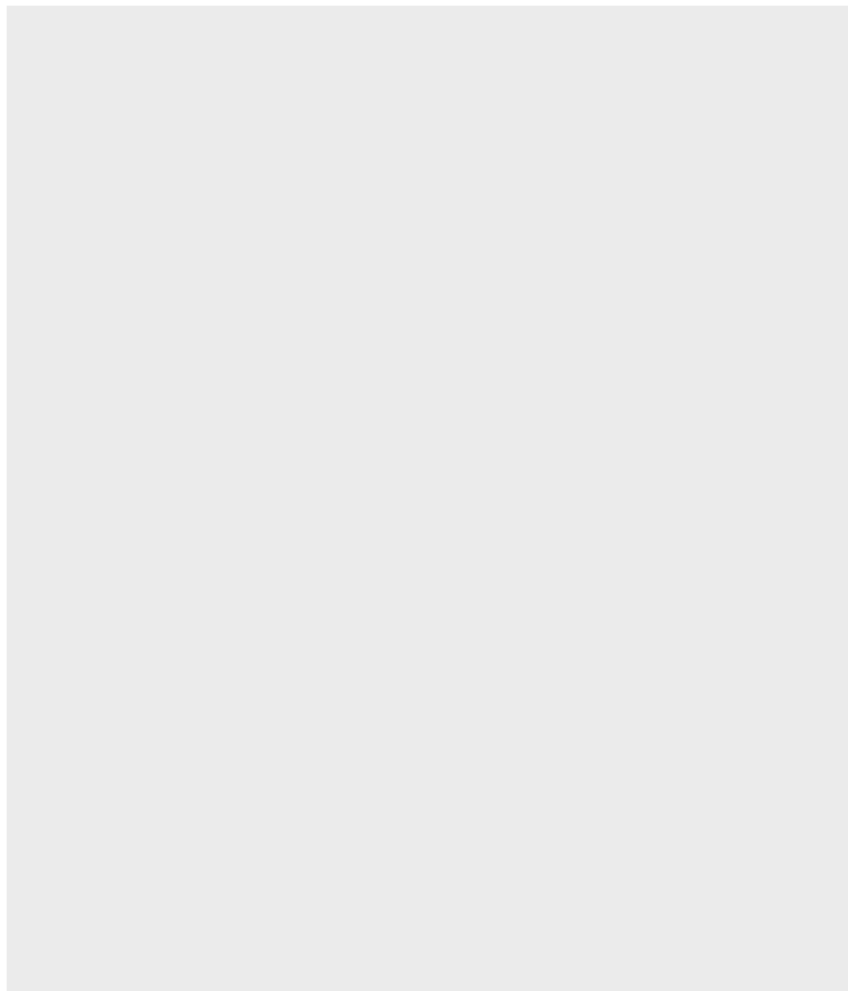
There are also different display modalities. Namely you can indicate if you want "both" the code and the output displayed in your flipbookification, or just the "output" (perhaps to be used in a traditional presentation), or just the "code" (which might be used to kind of test student expectations about some code). You have already seen the default where the parameter `display_type` is set to "both", but let's have a look at "output" and "code" only.

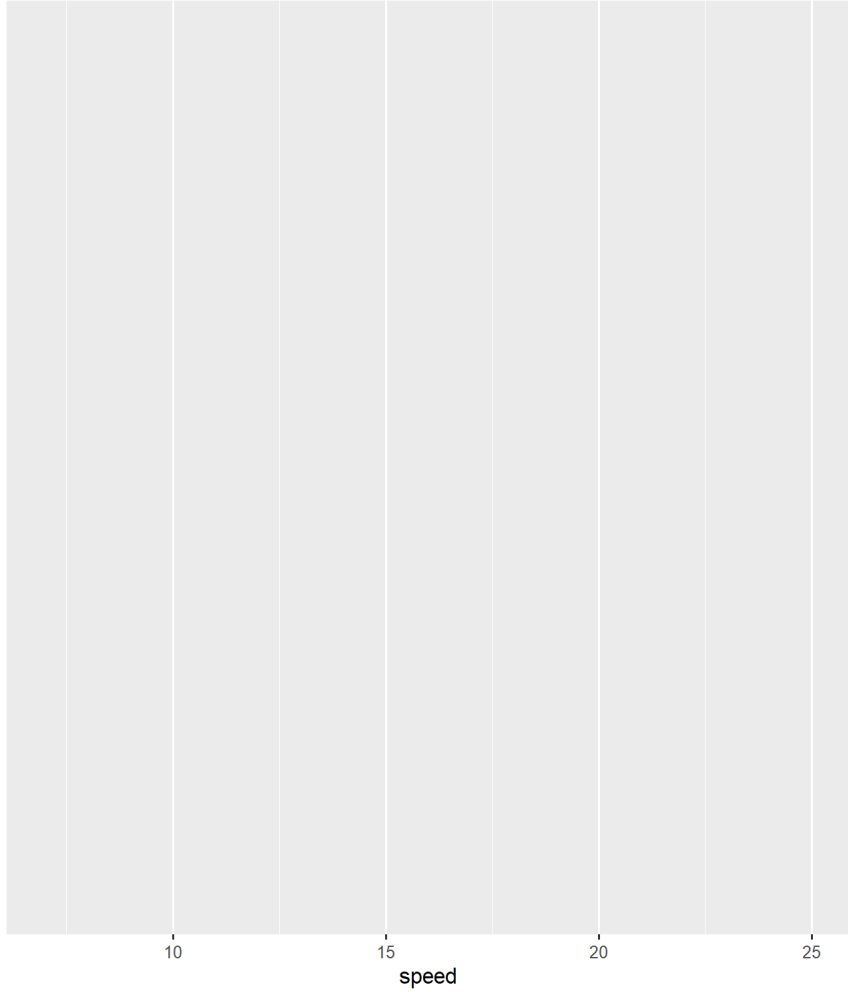
```
display_type = "output"
```

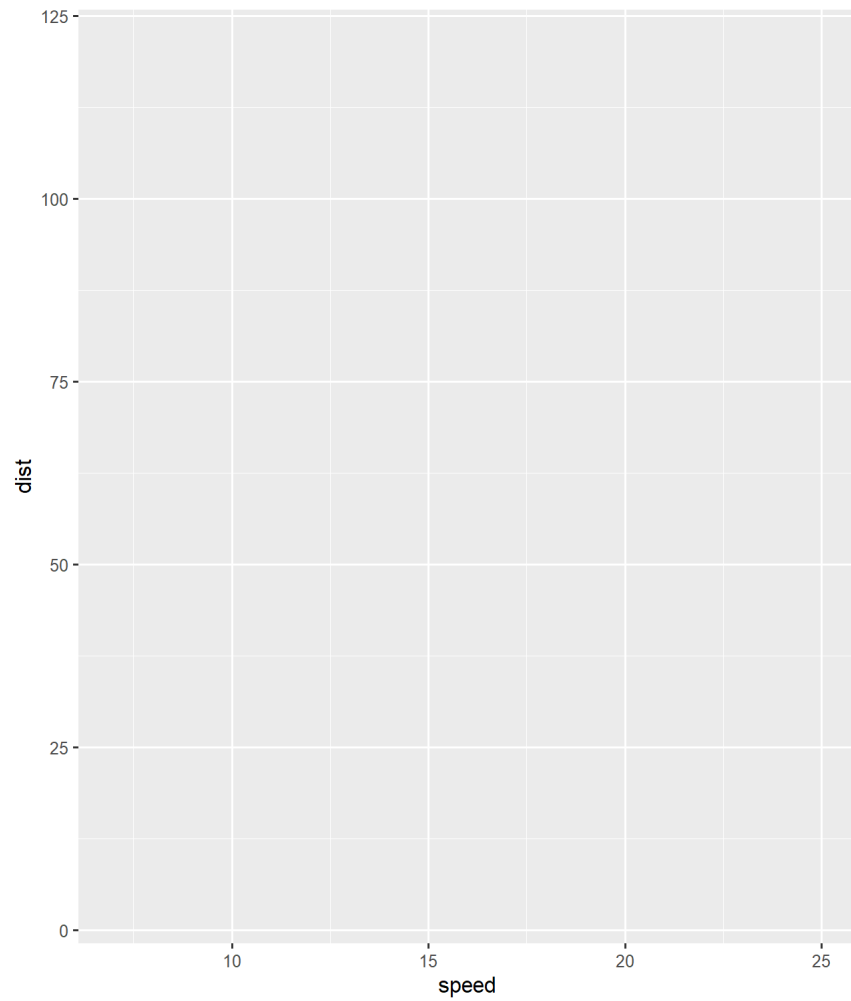
Let's look at where only the *output* is displayed for the "cars" code chunk.

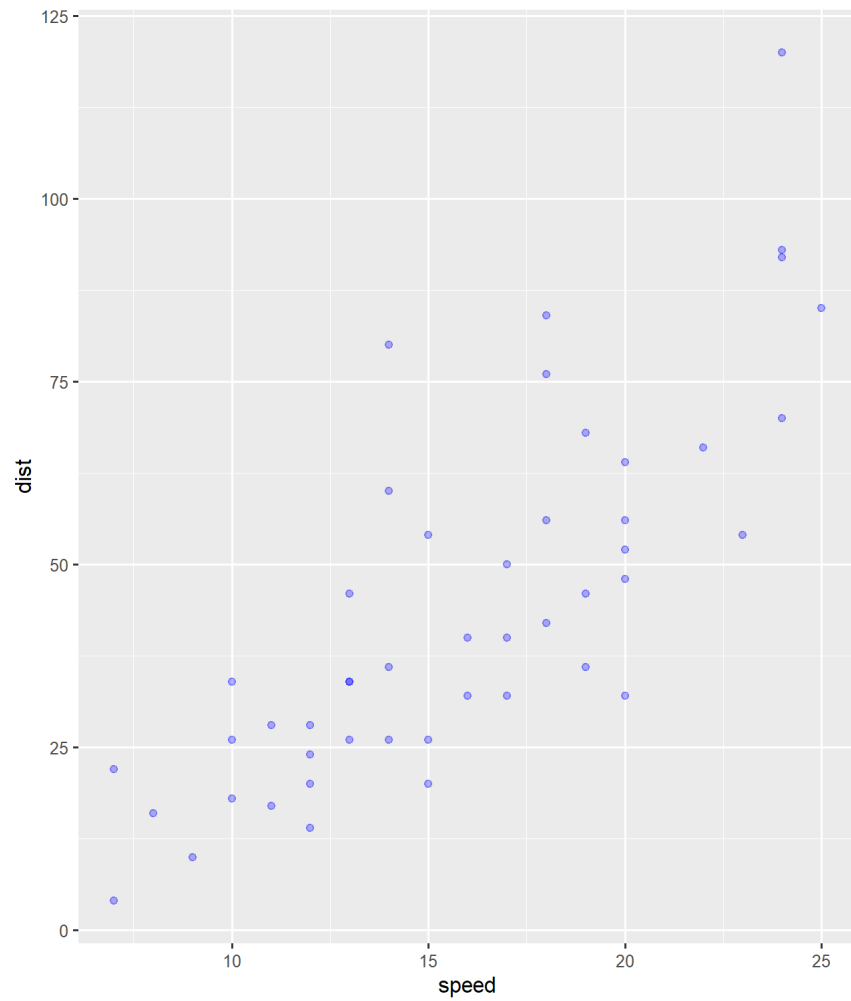
	speed	dist
1	4	2
2	4	10
3	7	4
4	7	22
5	8	16
6	9	10
7	10	18
8	10	26
9	10	34
10	11	17
11	11	28
12	12	14
13	12	20
14	12	24
15	12	28
16	13	26
17	13	34
18	13	34
19	13	46
20	14	26
21	14	36
22	14	60
23	14	80
24	15	20
25	15	26

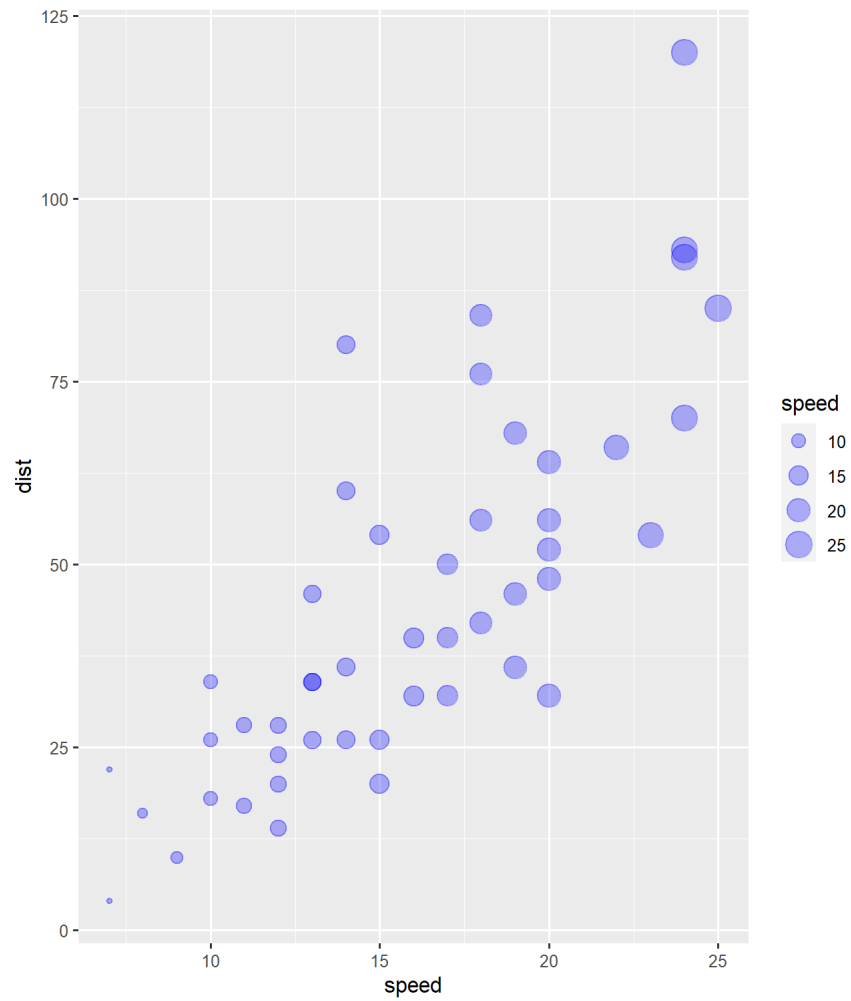
	speed	dist
1	7	4
2	7	22
3	8	16
4	9	10
5	10	18
6	10	26
7	10	34
8	11	17
9	11	28
10	12	14
11	12	20
12	12	24
13	12	28
14	13	26
15	13	34
16	13	34
17	13	46
18	14	26
19	14	36
20	14	60
21	14	80
22	15	20
23	15	26
24	15	54
25	16	32











```
display_type = "code"
```

And now where only the *code* is displayed for the "cars" code chunk.

`cars`

```
cars %>%
```

```
  filter(speed > 4)
```

```
cars %>%  
  filter(speed > 4) %>%  
  ggplot()
```

```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed)
```

```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist)
```

```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist) +  
  geom_point(  
    alpha = .3,  
    color = "blue"  
  )
```



```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist) +  
  geom_point(  
    alpha = .3,  
    color = "blue"  
  ) +  
  aes(size = speed)
```

```
display_type = c("output",  
"code")
```

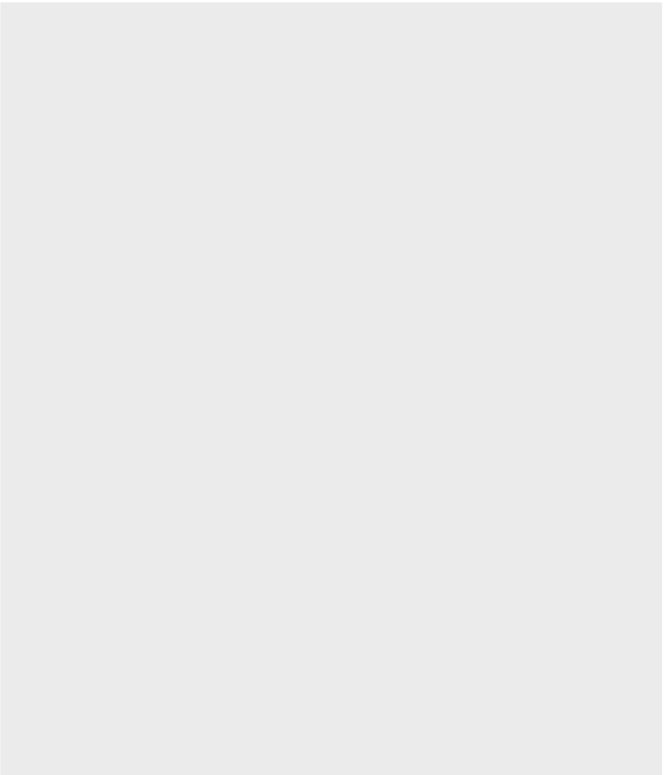
Not sure why you'd want to do this, but you can flip output and code. It's also not totally stable - jumps moving from tabular output to figure. Have to figure that one out. It is something to do with fixed height.

	speed	dist
1	4	2
2	4	10
3	7	4
4	7	22
5	8	16
6	9	10
7	10	18
8	10	26
9	10	34
10	11	17
11	11	28
12	12	14
13	12	20
14	12	24
15	12	28
16	13	26
17	13	34
18	13	34
19	13	46
20	14	26
21	14	36
22	14	60
23	14	80
24	15	20
25	15	26
26	15	54
27	16	32
28	16	40
29	17	32
30	17	40
31	17	50
32	18	42

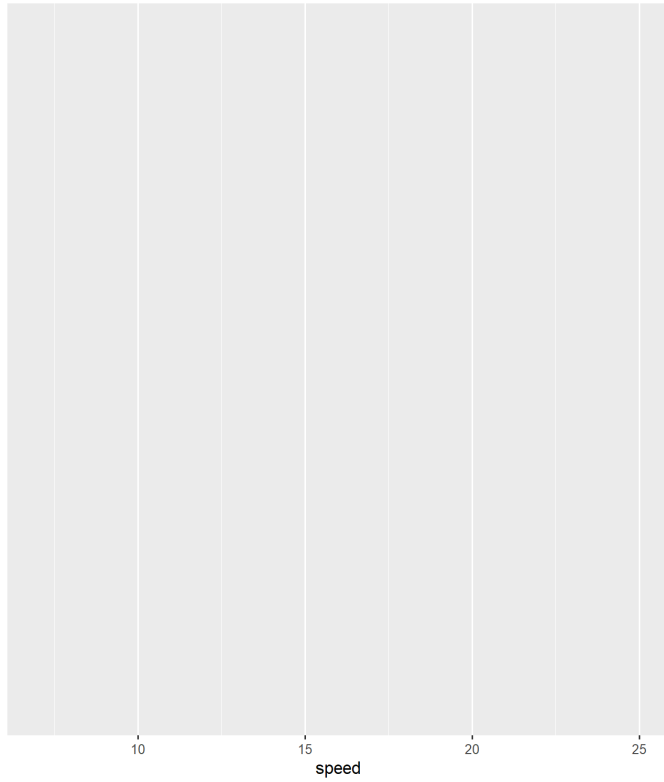
`cars`

	speed	dist
1	7	4
2	7	22
3	8	16
4	9	10
5	10	18
6	10	26
7	10	34
8	11	17
9	11	28
10	12	14
11	12	20
12	12	24
13	12	28
14	13	26
15	13	34
16	13	34
17	13	46
18	14	26
19	14	36
20	14	60
21	14	80
22	15	20
23	15	26
24	15	54
25	16	32
26	16	40
27	17	32
28	17	40
29	17	50
30	18	42
31	18	56
32	18	76

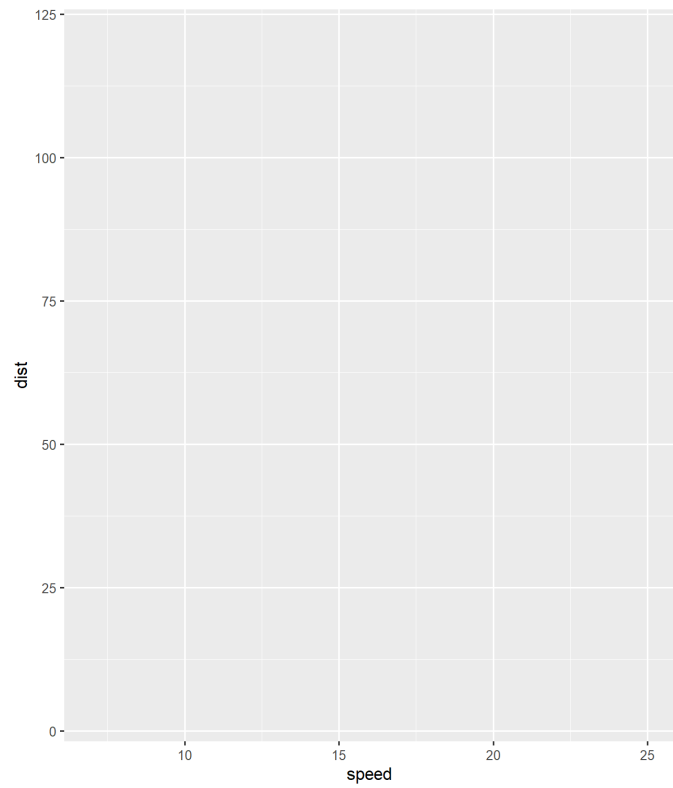
```
cars %>%  
  filter(speed > 4)
```



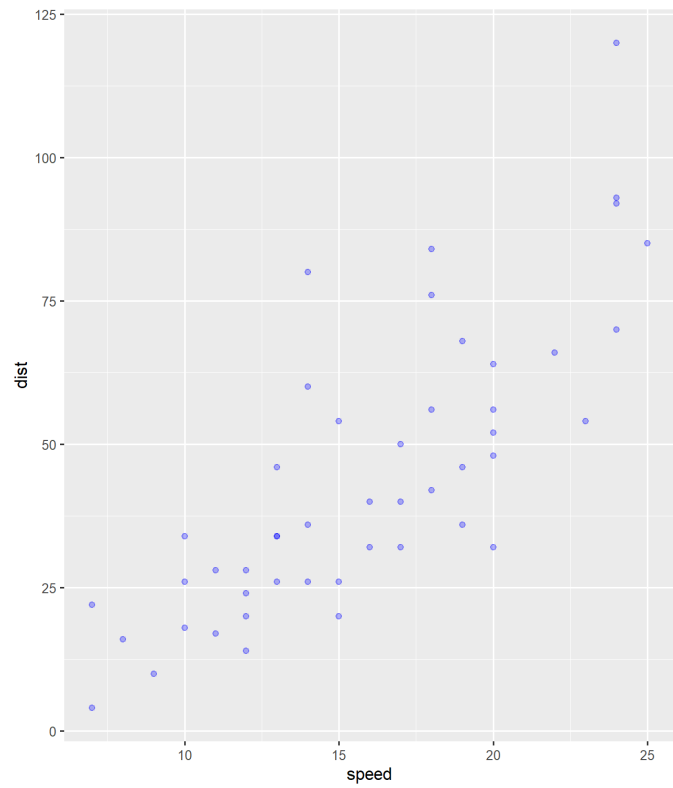
```
cars %>%  
  filter(speed > 4) %>%  
  ggplot()
```



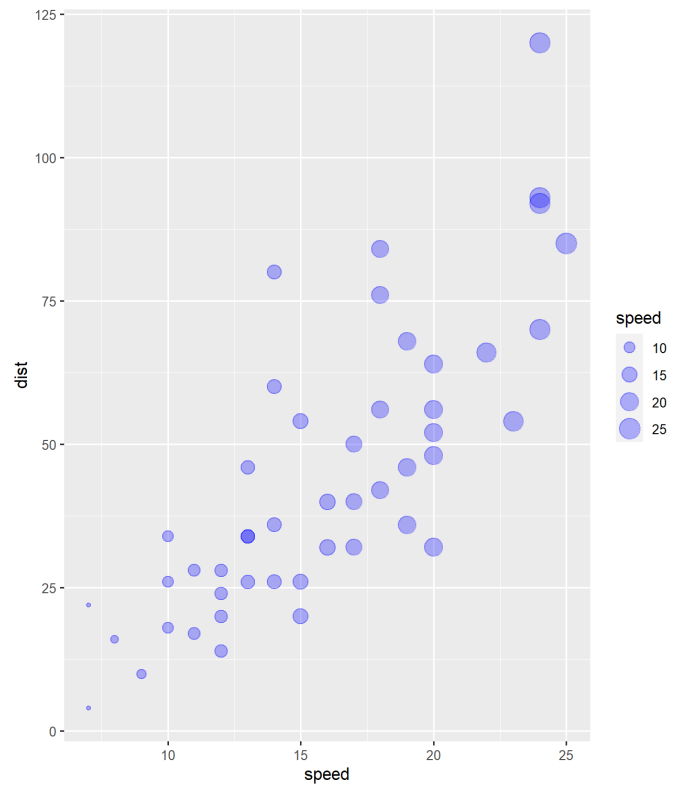
```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed)
```



```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist)
```



```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist) +  
  geom_point(  
    alpha = .3,  
    color = "blue"  
  )
```

```
cars %>%  
  filter(speed > 4) %>%  
  ggplot() +  
    aes(x = speed) +  
    aes(y = dist) +  
    geom_point(  
      alpha = .3,  
      color = "blue"  
    ) +  
    aes(size = speed)
```

Assignment

If you want to create an object in your flipbooks, it is most "natural" to use **right assignment**. Working sequentially with a pipeline of code, you get feedback all along the way until you get to the point of assigning all of what you have done to a new object with right assignment. Creating objects in one "source" code chunk, means that you can break up a pipeline of tasks into multiple code chunks. Let's see this in action.

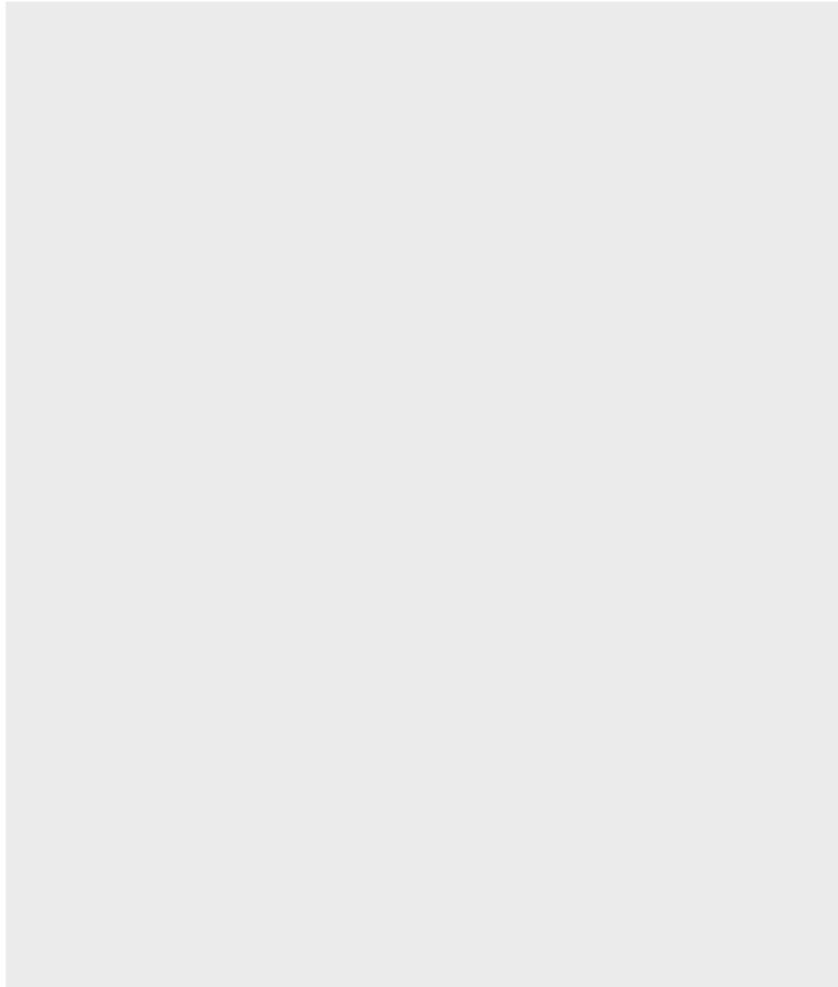
cars

speed dist

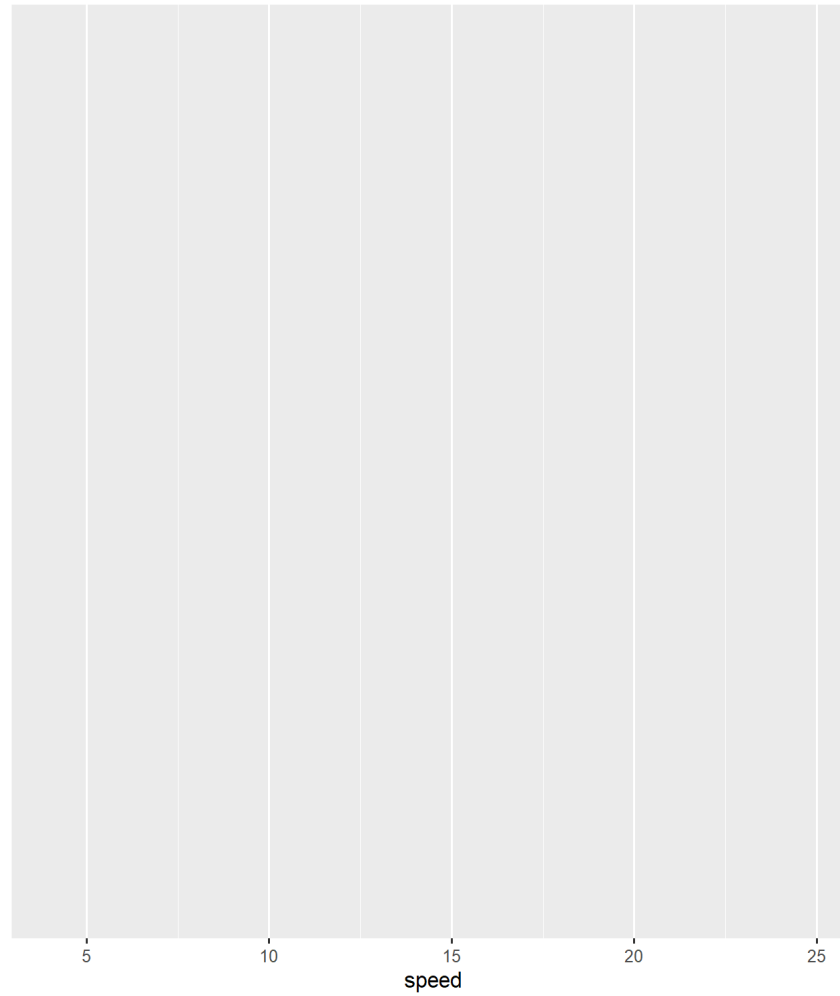
1	4	2
2	4	10
3	7	4
4	7	22
5	8	16
6	9	10
7	10	18
8	10	26
9	10	34
10	11	17
11	11	28
12	12	14
13	12	20
14	12	24
15	12	28
16	13	26
17	13	34
18	13	34
19	13	46
20	14	26
21	14	36
22	14	60
23	14	80
24	15	20
25	15	26
26	15	54
27	16	32
28	16	40
29	17	32
30	17	40
31	17	50
32	18	42

```
cars %>%
```

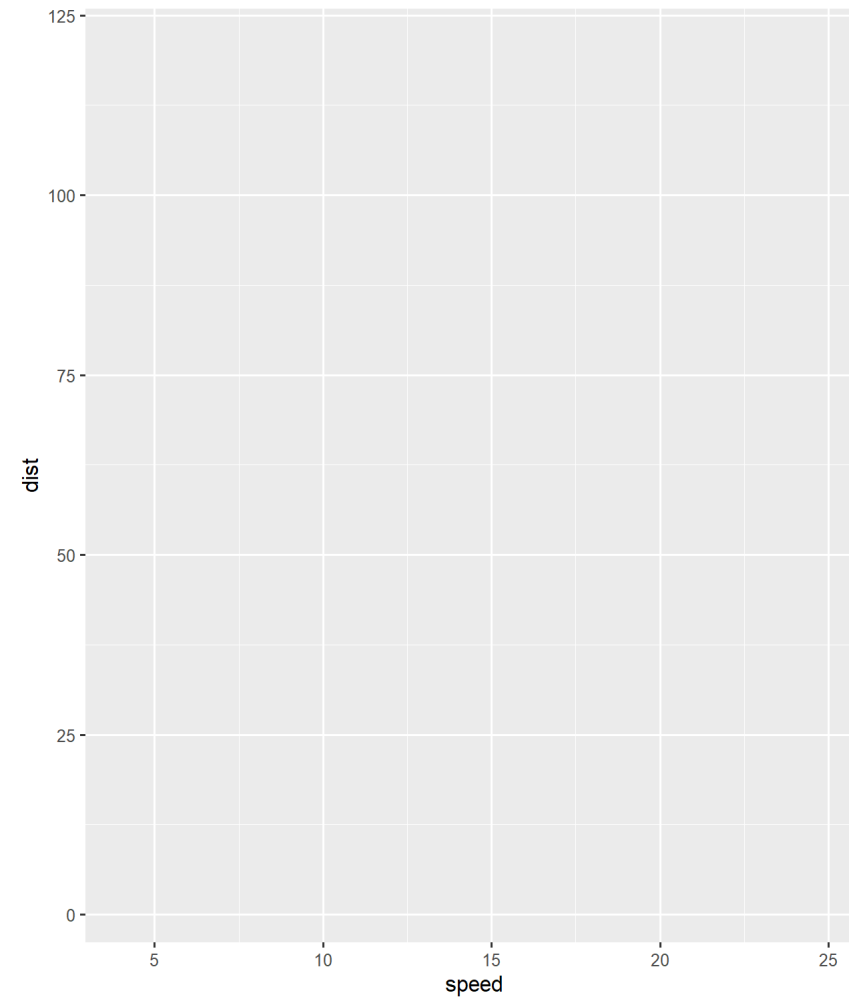
```
  ggplot()
```



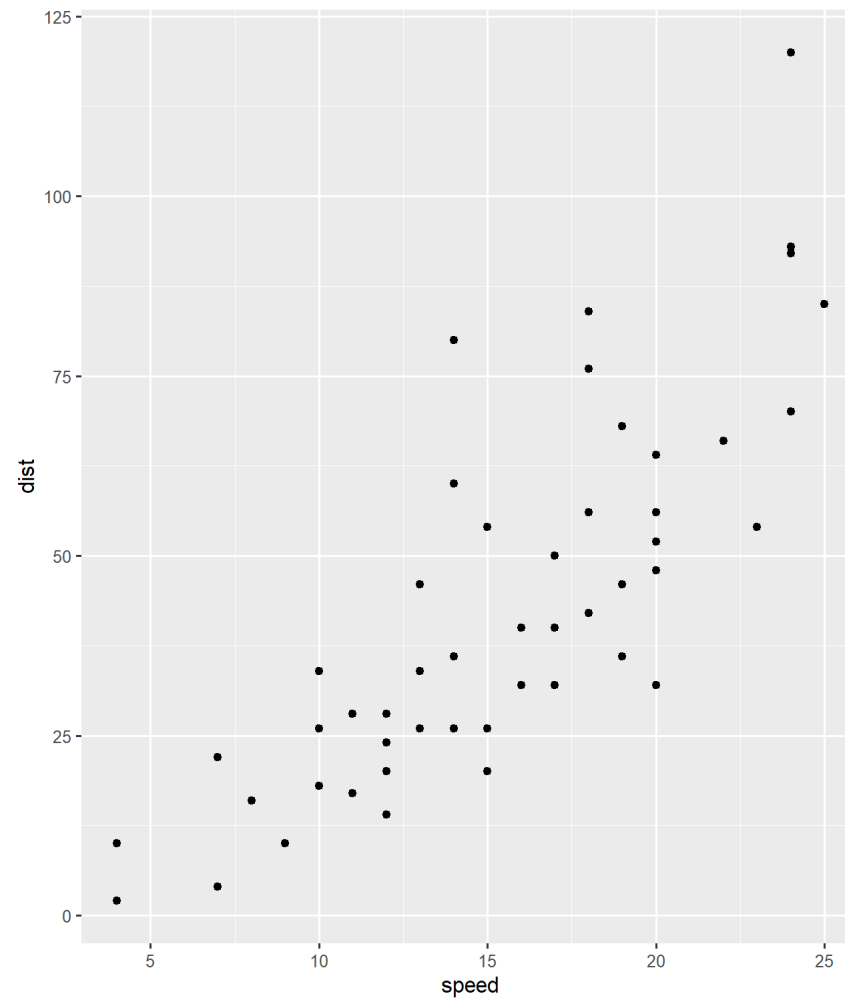
```
cars %>%  
  ggplot() +  
  aes(x = speed)
```



```
cars %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist)
```



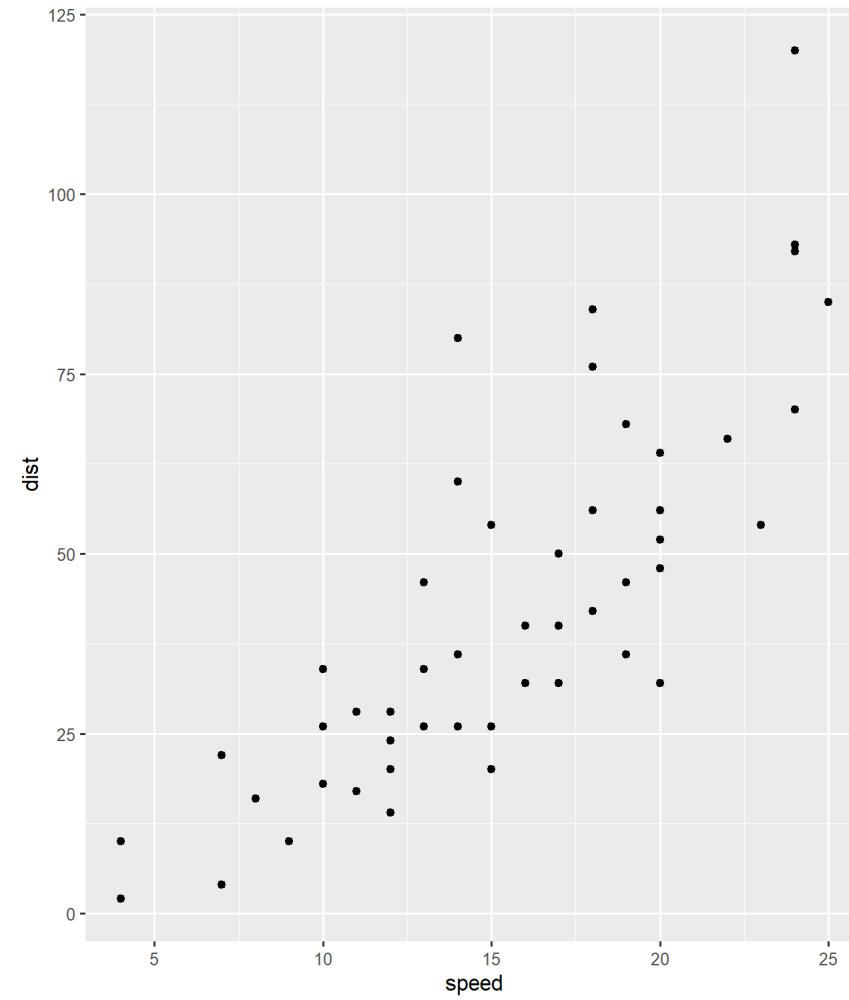
```
cars %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist) +  
  geom_point()
```



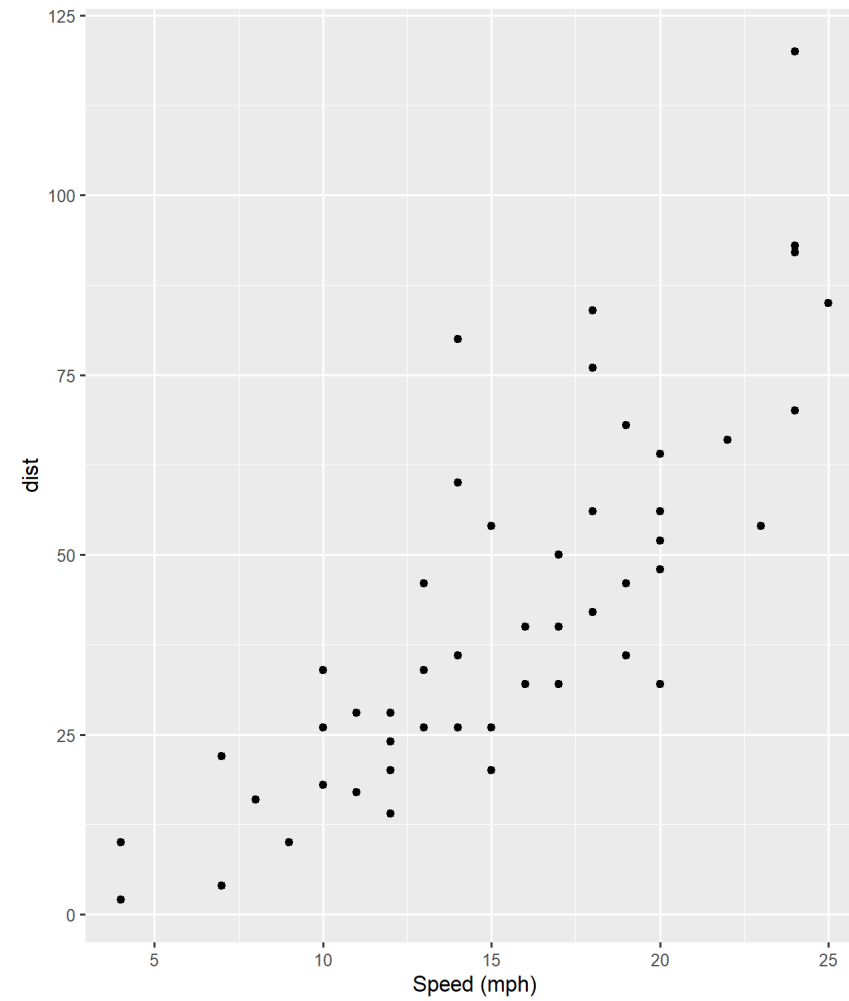
```
cars %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist) +  
  geom_point() ->  
cars_plot
```



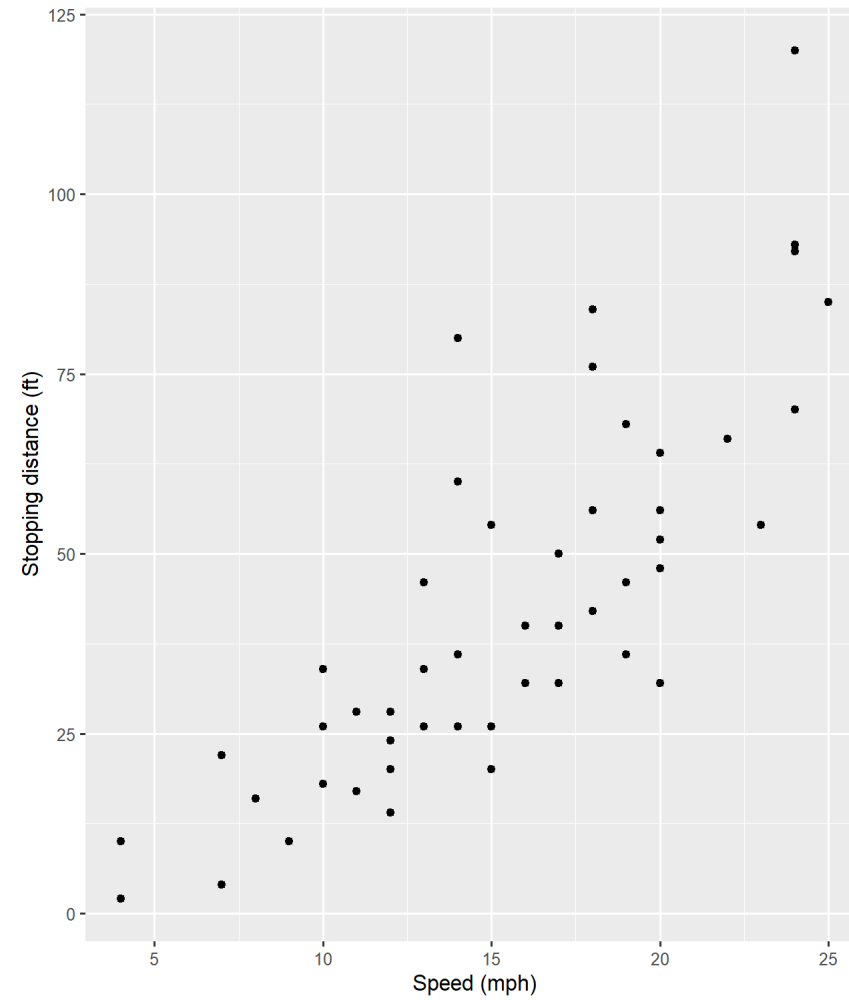
```
cars_plot
```



```
cars_plot +  
  labs(x = "Speed (mph)")
```



```
cars_plot +  
  labs(x = "Speed (mph)") +  
  labs(y = "Stopping distance (ft)")
```



`left_assign = TRUE`

With left assignment in R, you don't get any feedback, so flipbooking prefers this step at the end of a pipeline, so we can enjoy all the nice feedback. So the parameter `left_assign` is by default set to `FALSE`.

But, setting the `left_assign` parameter to `T` and using left assignment, you can still create a meaningful flipbook that gives you feedback. When `left_assign = TRUE`, the first object that is created prints at the end of the derivative code chunks.

```
my_plot <- cars # the data
```

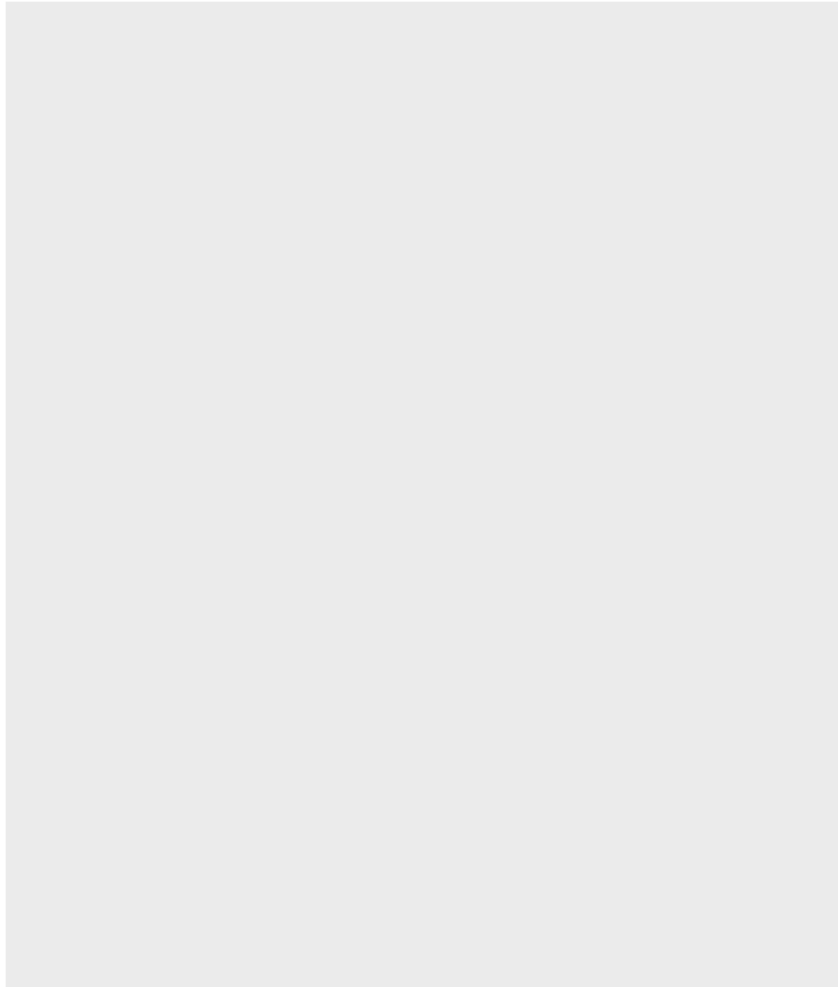
```
my_plot
```

	speed	dist
1	4	2
2	4	10
3	7	4
4	7	22
5	8	16
6	9	10
7	10	18
8	10	26
9	10	34
10	11	17
11	11	28
12	12	14
13	12	20
14	12	24
15	12	28
16	13	26
17	13	34
18	13	34
19	13	46
20	14	26
21	14	36
22	14	60
23	14	80
24	15	20
25	15	26
26	15	54
27	16	32
28	16	40
29	17	32
30	17	40
31	17	50
32	18	42

```
my_plot <- cars %>% # the data
  filter(speed > 4) # subset
my_plot
```

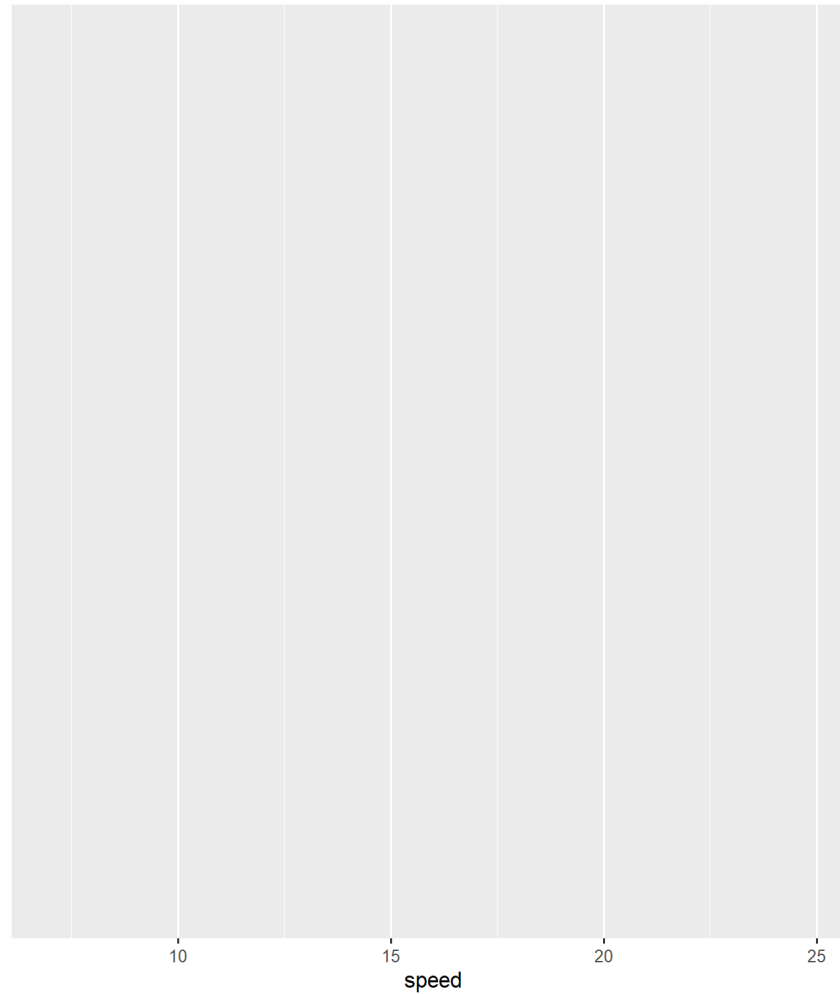
	speed	dist
1	7	4
2	7	22
3	8	16
4	9	10
5	10	18
6	10	26
7	10	34
8	11	17
9	11	28
10	12	14
11	12	20
12	12	24
13	12	28
14	13	26
15	13	34
16	13	34
17	13	46
18	14	26
19	14	36
20	14	60
21	14	80
22	15	20
23	15	26
24	15	54
25	16	32
26	16	40
27	17	32
28	17	40
29	17	50
30	18	42
31	18	56
32	18	76

```
my_plot <- cars %>% # the data
  filter(speed > 4) %>% # subset
  ggplot() # pipe to ggplot
my_plot
```

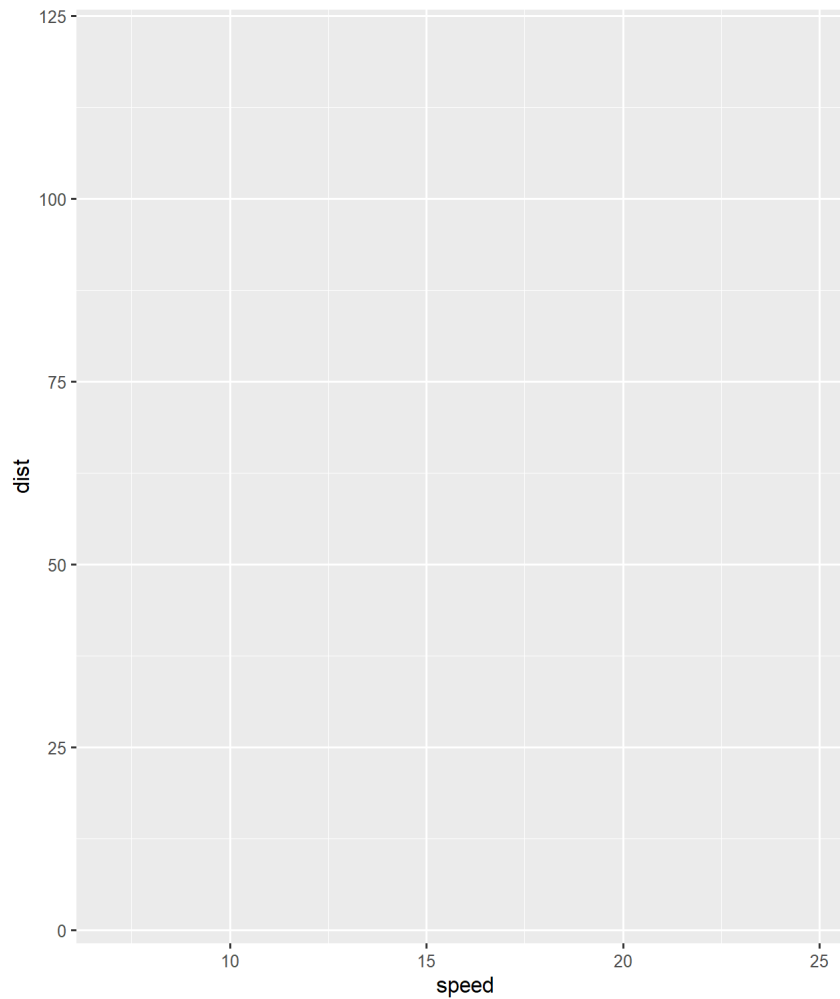


```
my_plot <- cars %>% # the data
  filter(speed > 4) %>% # subset
  ggplot() + # pipe to ggplot
  aes(x = speed)

my_plot
```

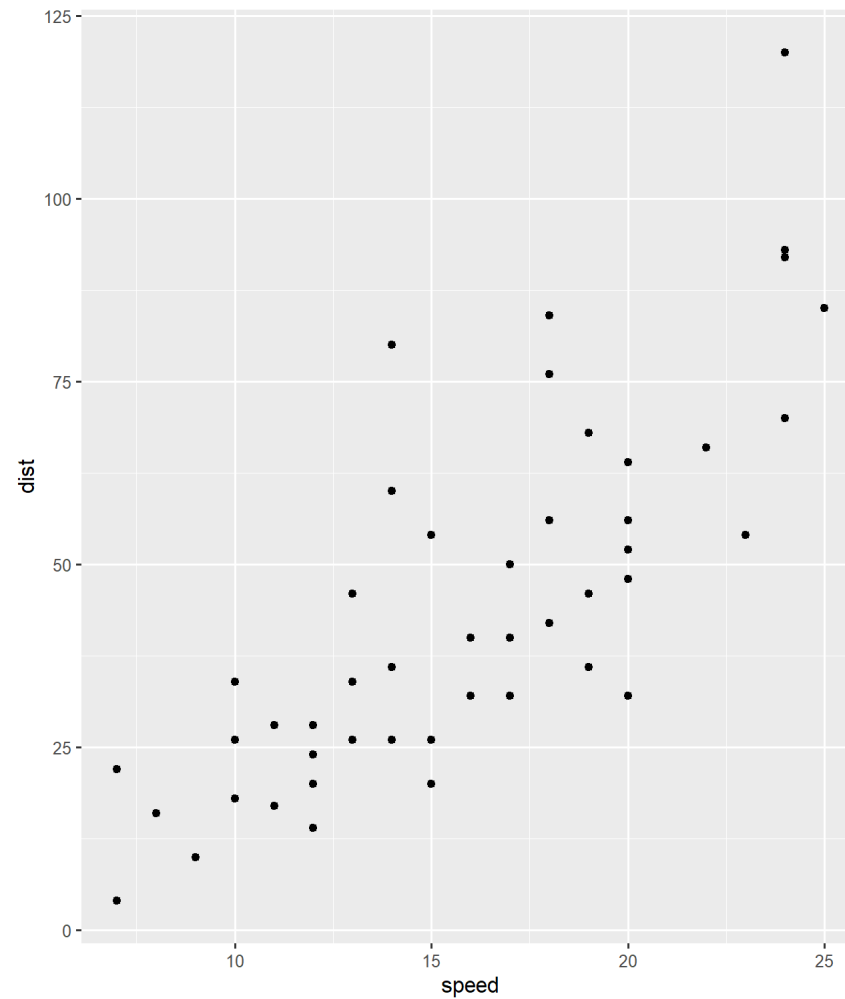



```
my_plot <- cars %>% # the data
  filter(speed > 4) %>% # subset
  ggplot() + # pipe to ggplot
  aes(x = speed) +
  aes(y = dist)
my_plot
```



```
my_plot <- cars %>% # the data
  filter(speed > 4) %>% # subset
  ggplot() + # pipe to ggplot
  aes(x = speed) +
  aes(y = dist) +
  geom_point()

my_plot
```



Managing source code chunks

So, it is pretty cool that we can create a bunch of derivative code chunks from one input code chunk (a foundational blog post by Emi Tanaka on this [here](#)). But there are some considerations then for this source chunk. What should its chunk options be? The easy way is to set all "source" code chunks to include = F, as I do throughout the book. However, you might consider a combination of `eval` and `echo` instead; you can come back to this idea as you become a more seasoned flipbooker.

Beyond the tidyverse

It is no surprise that Flipbooks are born in the context of the popularity of the tidyverse tools --- tools that are designed to be used in sequential pipelines and that give a satisfying amount of feedback along the way!

But base R techniques and other popular tools can certainly also be employed.

"chaining" by overwriting objects

```
cars_mod <- cars
```

```
cars_mod
```

	speed	dist
1	4	2
2	4	10
3	7	4
4	7	22
5	8	16
6	9	10
7	10	18
8	10	26
9	10	34
10	11	17
11	11	28
12	12	14
13	12	20
14	12	24
15	12	28
16	13	26
17	13	34
18	13	34
19	13	46
20	14	26
21	14	36
22	14	60
23	14	80
24	15	20
25	15	26
26	15	54
27	16	32
28	16	40
29	17	32
30	17	40
31	17	50
32	18	42

```
cars_mod <- cars
```

```
cars_mod$half_dist <- cars$dist / 2
```

```
cars_mod
```

	speed	dist	half_dist
1	4	2	1.0
2	4	10	5.0
3	7	4	2.0
4	7	22	11.0
5	8	16	8.0
6	9	10	5.0
7	10	18	9.0
8	10	26	13.0
9	10	34	17.0
10	11	17	8.5
11	11	28	14.0
12	12	14	7.0
13	12	20	10.0
14	12	24	12.0
15	12	28	14.0
16	13	26	13.0
17	13	34	17.0
18	13	34	17.0
19	13	46	23.0
20	14	26	13.0
21	14	36	18.0
22	14	60	30.0
23	14	80	40.0
24	15	20	10.0
25	15	26	13.0
26	15	54	27.0
27	16	32	16.0
28	16	40	20.0
29	17	32	16.0
30	17	40	20.0
31	17	50	25.0
32	18	42	21.0

```
cars_mod <- cars
cars_mod$half_dist <- cars$dist / 2
names(cars_mod)[2] <- "distance"

cars_mod
```

	speed	distance	half_dist
1	4	2	1.0
2	4	10	5.0
3	7	4	2.0
4	7	22	11.0
5	8	16	8.0
6	9	10	5.0
7	10	18	9.0
8	10	26	13.0
9	10	34	17.0
10	11	17	8.5
11	11	28	14.0
12	12	14	7.0
13	12	20	10.0
14	12	24	12.0
15	12	28	14.0
16	13	26	13.0
17	13	34	17.0
18	13	34	17.0
19	13	46	23.0
20	14	26	13.0
21	14	36	18.0
22	14	60	30.0
23	14	80	40.0
24	15	20	10.0
25	15	26	13.0
26	15	54	27.0
27	16	32	16.0
28	16	40	20.0
29	17	32	16.0
30	17	40	20.0
31	17	50	25.0
32	18	42	21.0


```
cars_mod <- cars
cars_mod$half_dist <- cars$dist / 2
names(cars_mod)[2] <- "distance"
cars_mod <- cars_mod[cars_mod$distance > 10,]

cars_mod
```

	speed	distance	half_dist
4	7	22	11.0
5	8	16	8.0
7	10	18	9.0
8	10	26	13.0
9	10	34	17.0
10	11	17	8.5
11	11	28	14.0
12	12	14	7.0
13	12	20	10.0
14	12	24	12.0
15	12	28	14.0
16	13	26	13.0
17	13	34	17.0
18	13	34	17.0
19	13	46	23.0
20	14	26	13.0
21	14	36	18.0
22	14	60	30.0
23	14	80	40.0
24	15	20	10.0
25	15	26	13.0
26	15	54	27.0
27	16	32	16.0
28	16	40	20.0
29	17	32	16.0
30	17	40	20.0
31	17	50	25.0
32	18	42	21.0
33	18	56	28.0
34	18	76	38.0
35	18	84	42.0
36	19	36	18.0

```
cars_mod <- cars
cars_mod$half_dist <- cars$dist / 2
names(cars_mod)[2] <- "distance"
cars_mod <- cars_mod[cars_mod$distance > 10,]
cars_mod <- cars_mod["distance"]

cars_mod
```

	distance
4	22
5	16
7	18
8	26
9	34
10	17
11	28
12	14
13	20
14	24
15	28
16	26
17	34
18	34
19	46
20	26
21	36
22	60
23	80
24	20
25	26
26	54
27	32
28	40
29	32
30	40
31	50
32	42
33	56
34	76
35	84
36	36

using the `.[]` and `.[[[]]` syntax with the `migritr` pipe - `%>%`

Flipbooking can also be applied to logical indexing workflows if the steps are broken up using the `%>%` followed by `.[]` and `.[[[]]`. Thus flipbooking can also be used with base R logical indexing and with the popular `data.table` package.

cars

speed dist

1	4	2
2	4	10
3	7	4
4	7	22
5	8	16
6	9	10
7	10	18
8	10	26
9	10	34
10	11	17
11	11	28
12	12	14
13	12	20
14	12	24
15	12	28
16	13	26
17	13	34
18	13	34
19	13	46
20	14	26
21	14	36
22	14	60
23	14	80
24	15	20
25	15	26
26	15	54
27	16	32
28	16	40
29	17	32
30	17	40
31	17	50
32	18	42

```
cars %>%
```

```
  .[cars$speed > median(cars$speed),]
```

	speed	dist
27	16	32
28	16	40
29	17	32
30	17	40
31	17	50
32	18	42
33	18	56
34	18	76
35	18	84
36	19	36
37	19	46
38	19	68
39	20	32
40	20	48
41	20	52
42	20	56
43	20	64
44	22	66
45	23	54
46	24	70
47	24	92
48	24	93
49	24	120
50	25	85

```
cars %>%  
  .[cars$speed > median(cars$speed),] %>%  
  .["speed"]
```

	speed
27	16
28	16
29	17
30	17
31	17
32	18
33	18
34	18
35	18
36	19
37	19
38	19
39	20
40	20
41	20
42	20
43	20
44	22
45	23
46	24
47	24
48	24
49	24
50	25

```
cars %>%  
  .[cars$speed > median(cars$speed),] %>%  
  .["speed"] %>%  
  .[,1]
```

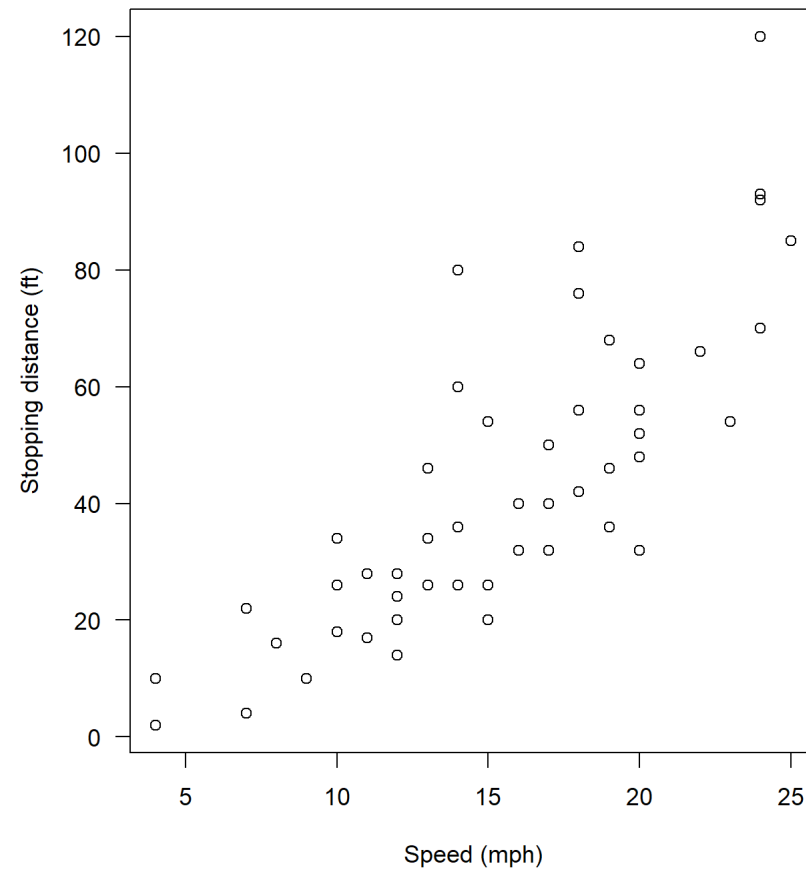
```
[1] 16 16 17 17 17 18 18 18 18 19 19 19 20 20 20 20 20 22 23 24 24 24 24 25
```

```
cars %>%  
  .[cars$speed > median(cars$speed),] %>%  
  .["speed"] %>%  
  .[,1] ->  
top_speeds
```

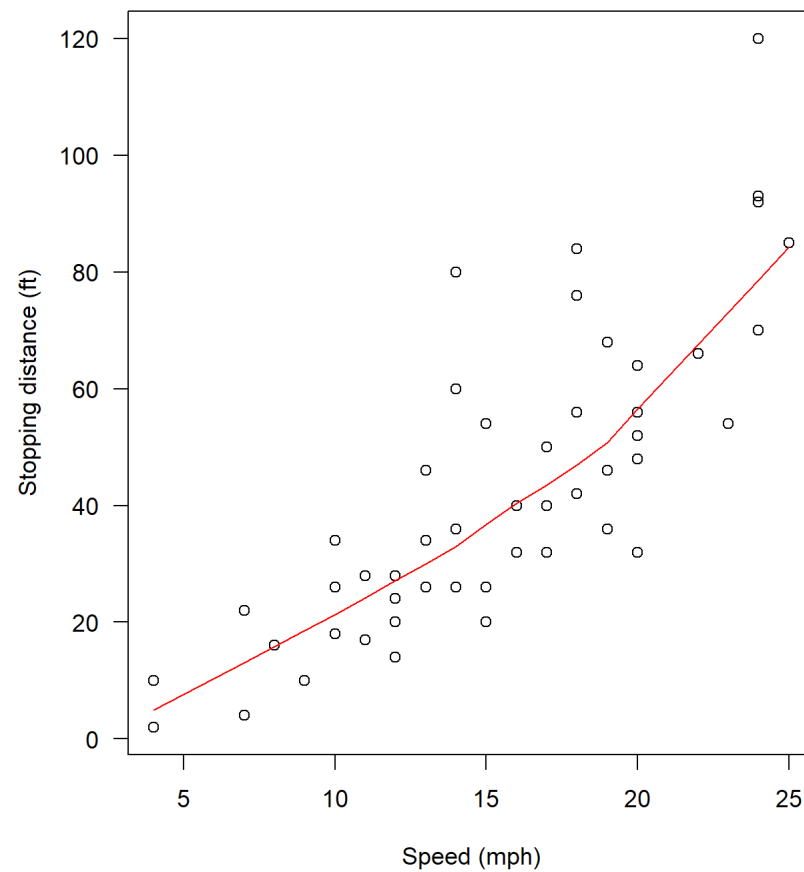

Base R plotting

It has been a while since I've done much plotting with base R, but I think it is important to have an example or two.

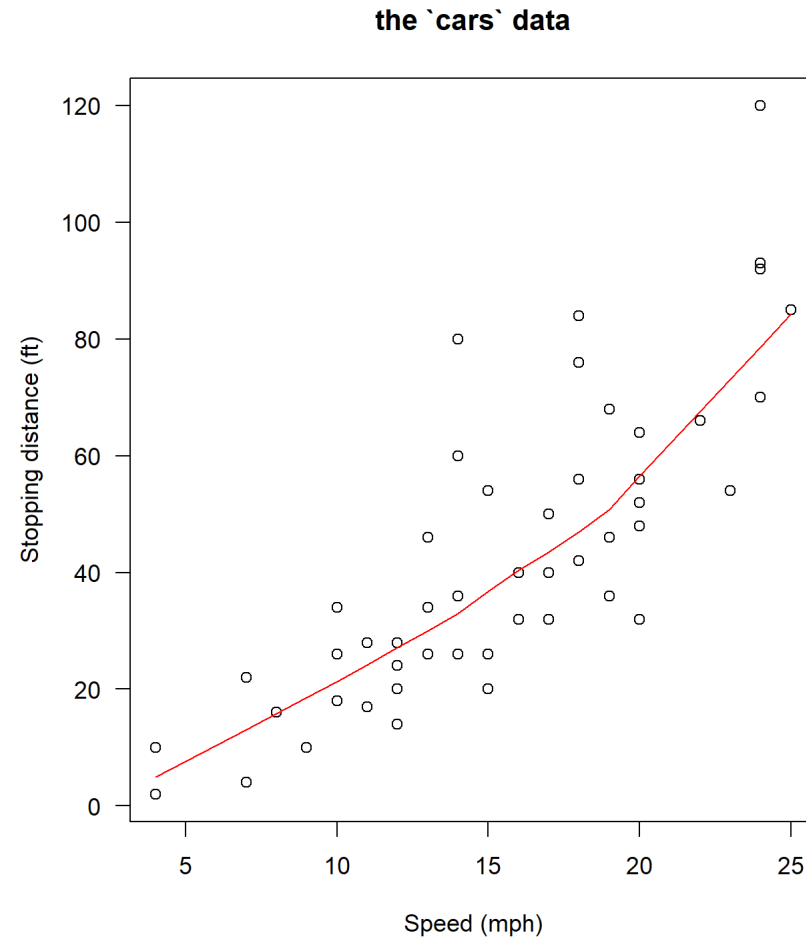
```
plot(cars, xlab = "Speed (mph)",  
      ylab = "Stopping distance (ft)",  
      las = 1)
```



```
plot(cars, xlab = "Speed (mph)",  
      ylab = "Stopping distance (ft)",  
      las = 1)  
lines(lowess(cars$speed, cars$dist,  
             f = 2/3, iter = 3),  
      col = "red")
```



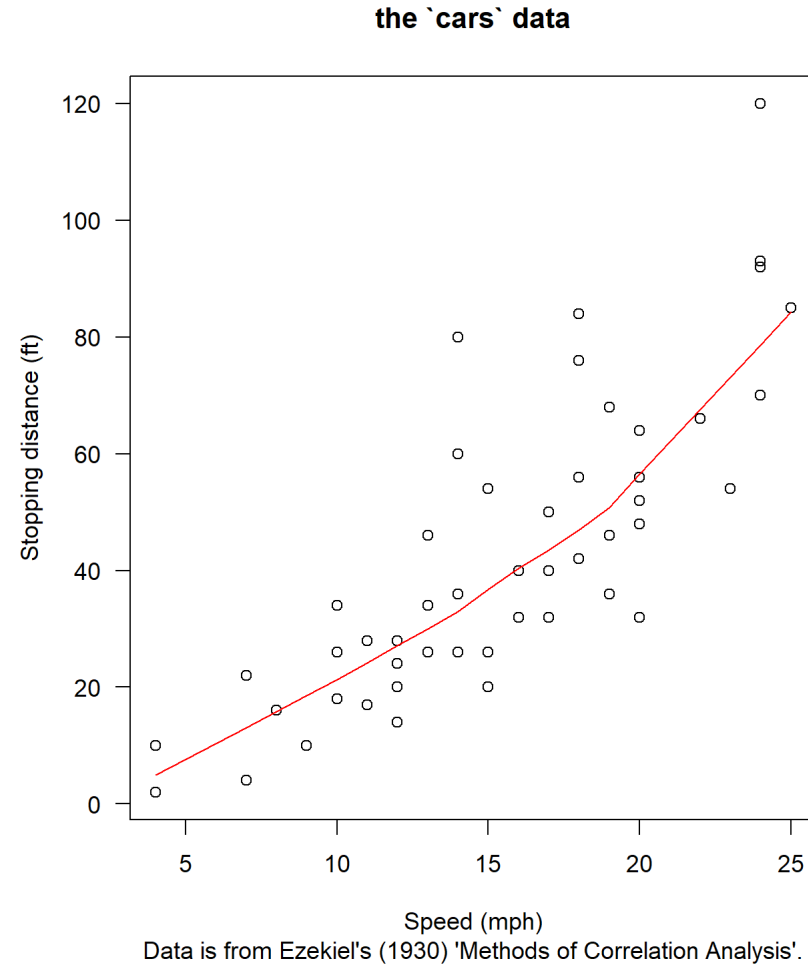
```
plot(cars, xlab = "Speed (mph)",  
     ylab = "Stopping distance (ft)",  
     las = 1)  
lines(lowess(cars$speed, cars$dist,  
            f = 2/3, iter = 3),  
      col = "red")  
title(main = "the `cars` data")
```



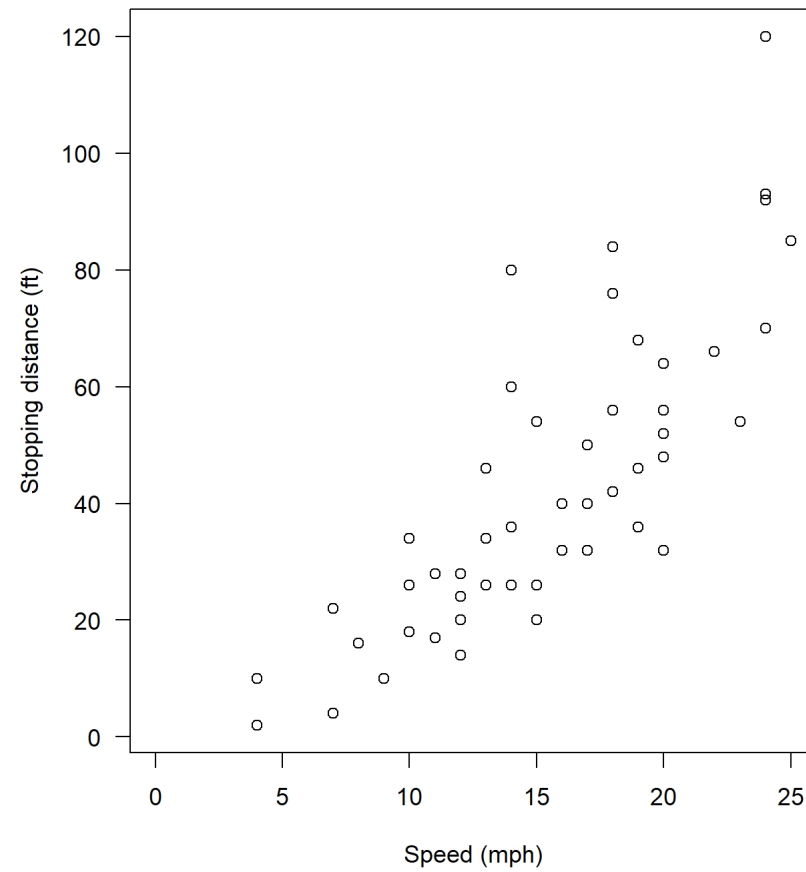
```

plot(cars, xlab = "Speed (mph)",
      ylab = "Stopping distance (ft)",
      las = 1)
lines(lowess(cars$speed, cars$dist,
             f = 2/3, iter = 3),
      col = "red")
title(main = "the `cars` data")
title(sub = "Data is from Ezekiel's (1930) 'Methods of Correlation Analysis'")

```

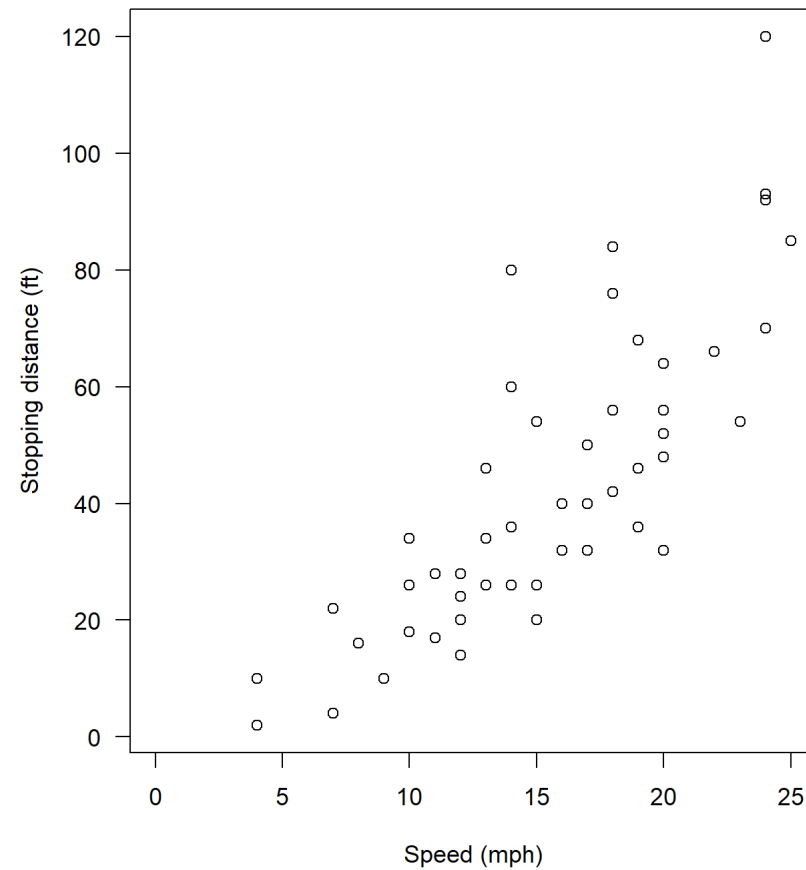


```
## An example of polynomial regression  
plot(cars, xlab = "Speed (mph)",  
      ylab = "Stopping distance (ft)",  
      las = 1, xlim = c(0, 25))
```



```
## An example of polynomial regression
plot(cars, xlab = "Speed (mph)",
      ylab = "Stopping distance (ft)",
      las = 1, xlim = c(0, 25))
```

```
lm(dist ~ poly(speed, 3),
   data = cars)
```



Call:

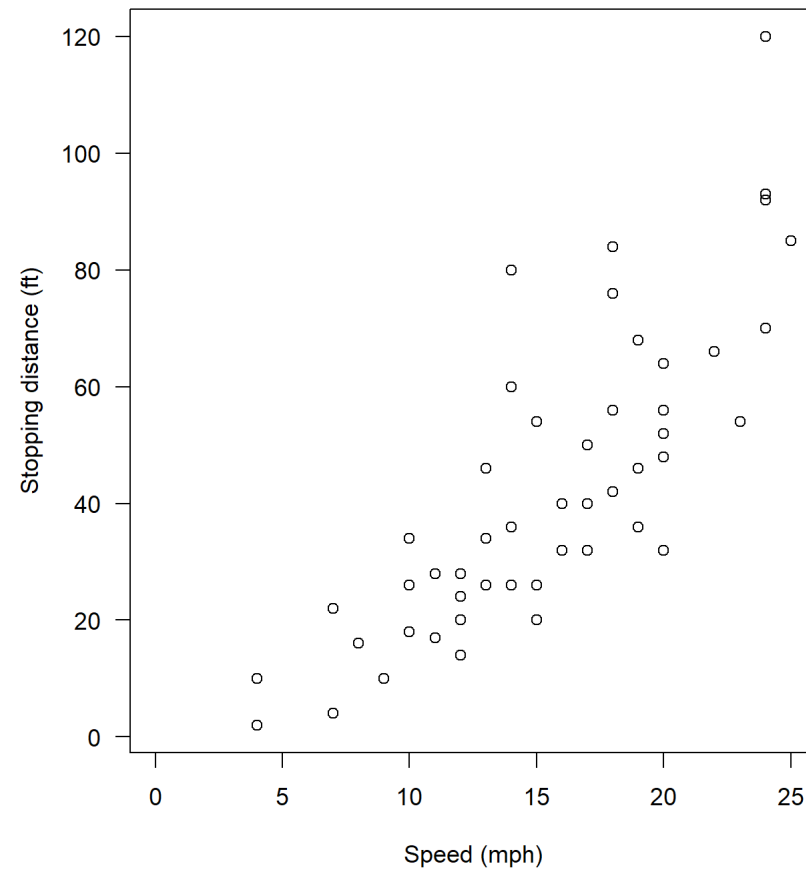
```
lm(formula = dist ~ poly(speed, 3), data = cars)
```

Coefficients:

(Intercept)	poly(speed, 3)1	poly(speed, 3)2	poly(speed, 3)3
42.98	145.55	23.00	13.80

```
## An example of polynomial regression
plot(cars, xlab = "Speed (mph)",
      ylab = "Stopping distance (ft)",
      las = 1, xlim = c(0, 25))

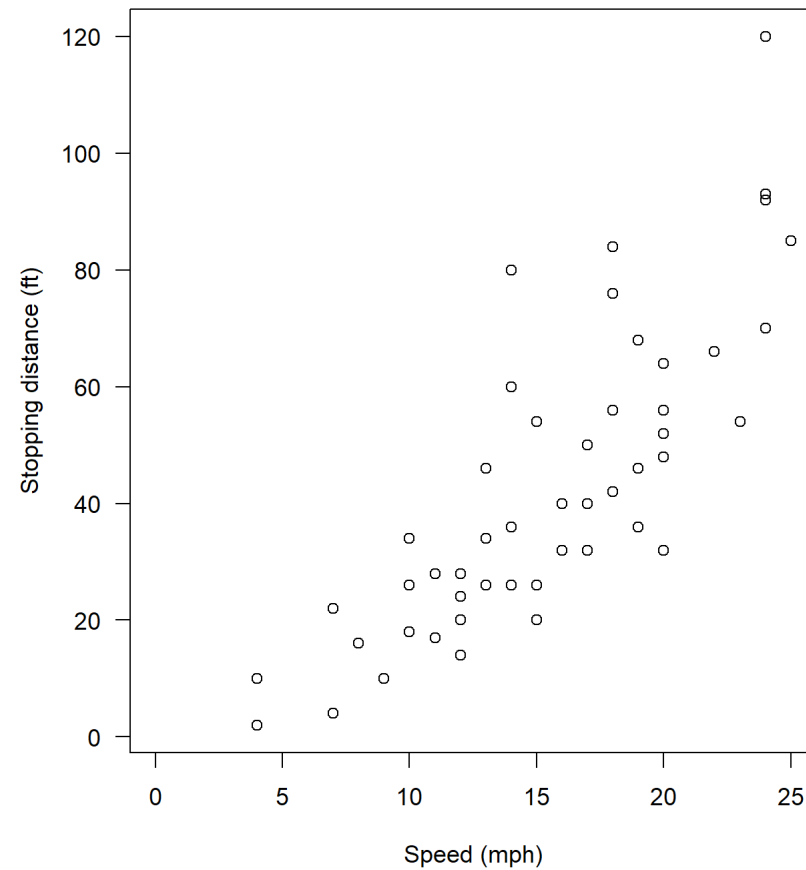
lm(dist ~ poly(speed, 3),
  data = cars) ->
model
```




```
## An example of polynomial regression
plot(cars, xlab = "Speed (mph)",
      ylab = "Stopping distance (ft)",
      las = 1, xlim = c(0, 25))

lm(dist ~ poly(speed, 3),
    data = cars) ->
model

seq(0, 25, length.out = 25)
```

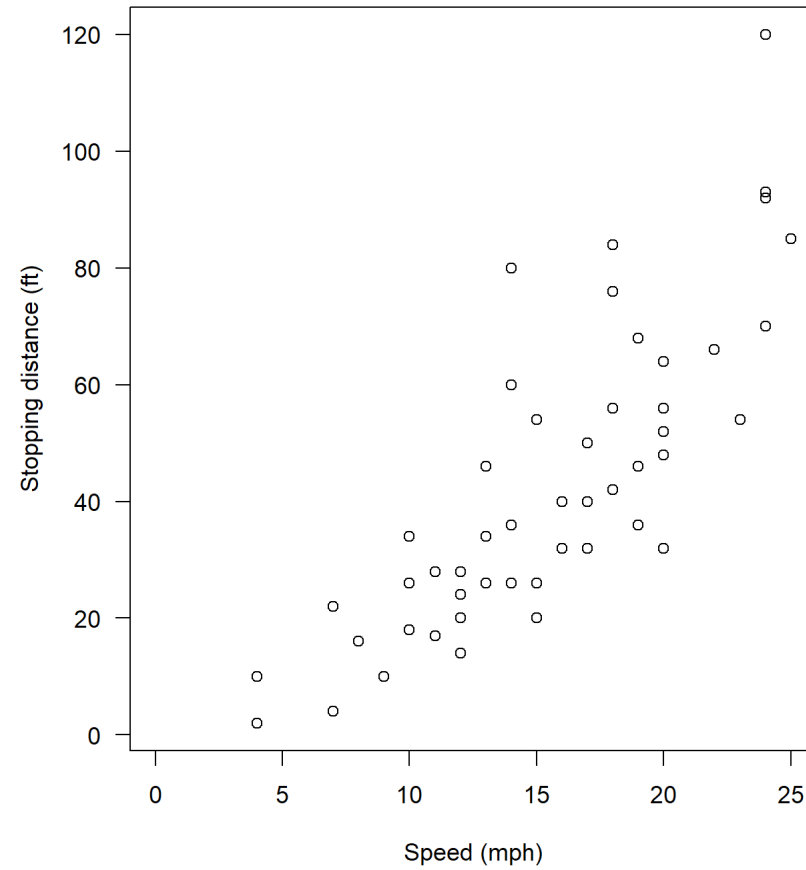


```
[1] 0.000000 1.041667 2.083333 3.125000 4.166667 5.208333 6.250000 7.29166
[9] 8.333333 9.375000 10.416667 11.458333 12.500000 13.541667 14.583333 15.62500
[17] 16.666667 17.708333 18.750000 19.791667 20.833333 21.875000 22.916667 23.95833
[25] 25.000000
```

```
## An example of polynomial regression
plot(cars, xlab = "Speed (mph)",
      ylab = "Stopping distance (ft)",
      las = 1, xlim = c(0, 25))

lm(dist ~ poly(speed, 3),
    data = cars) ->
model

seq(0, 25, length.out = 25) ->
inputs_of_x
```

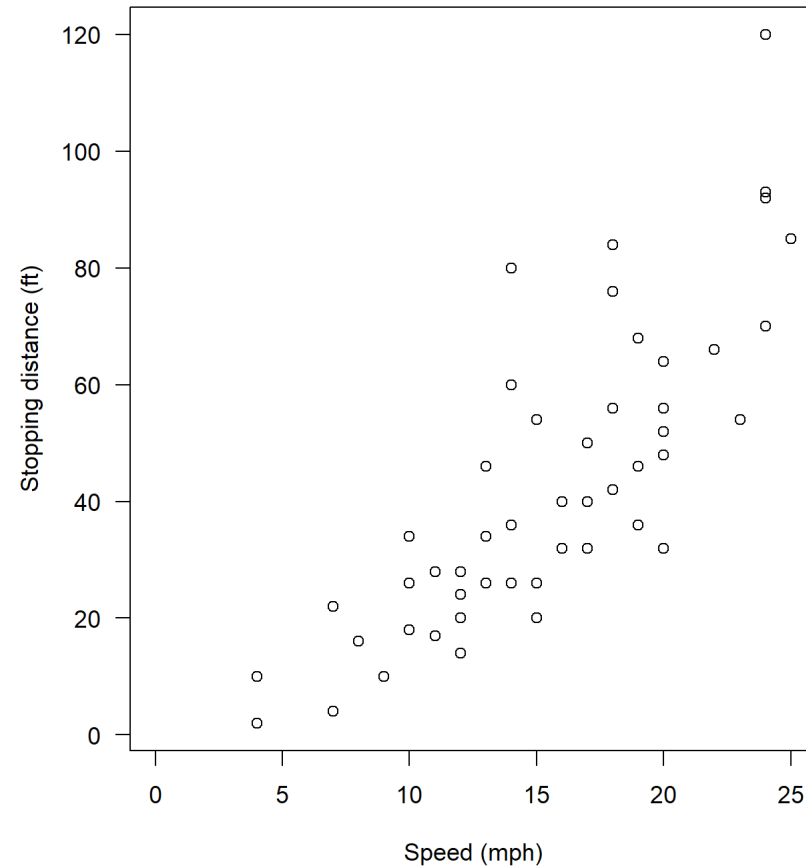


```
## An example of polynomial regression
plot(cars, xlab = "Speed (mph)",
     ylab = "Stopping distance (ft)",
     las = 1, xlim = c(0, 25))

lm(dist ~ poly(speed, 3),
   data = cars) ->
model

seq(0, 25, length.out = 25) ->
inputs_of_x

predict(model,
        data.frame(speed = inputs_of_x))
```



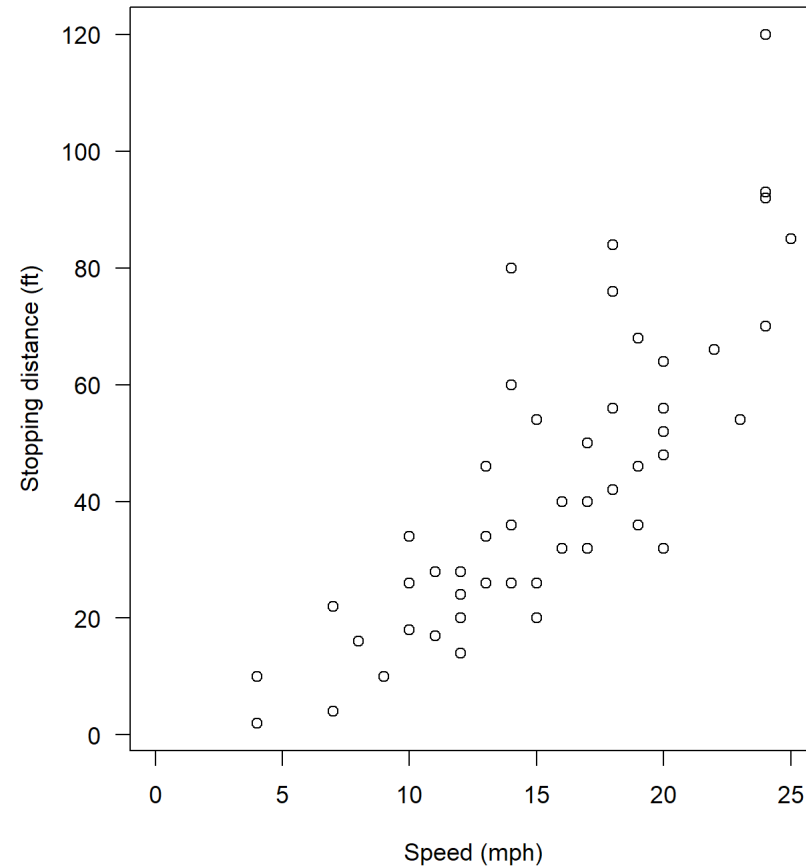
1	2	3	4	5	6	7
-19.505049	-12.788379	-6.760989	-1.353352	3.504057	7.880764	11.846296
8	9	10	11	12	13	14
15.470179	18.821939	21.971103	24.987195	27.939743	30.898273	33.932311
15	16	17	18	19	20	21
37.111382	40.505014	44.182731	48.214061	52.668530	57.615663	63.124987

```
## An example of polynomial regression
plot(cars, xlab = "Speed (mph)",
      ylab = "Stopping distance (ft)",
      las = 1, xlim = c(0, 25))

lm(dist ~ poly(speed, 3),
   data = cars) ->
model

seq(0, 25, length.out = 25) ->
inputs_of_x

predict(model,
        data.frame(speed = inputs_of_x)) ->
prediction_y
```



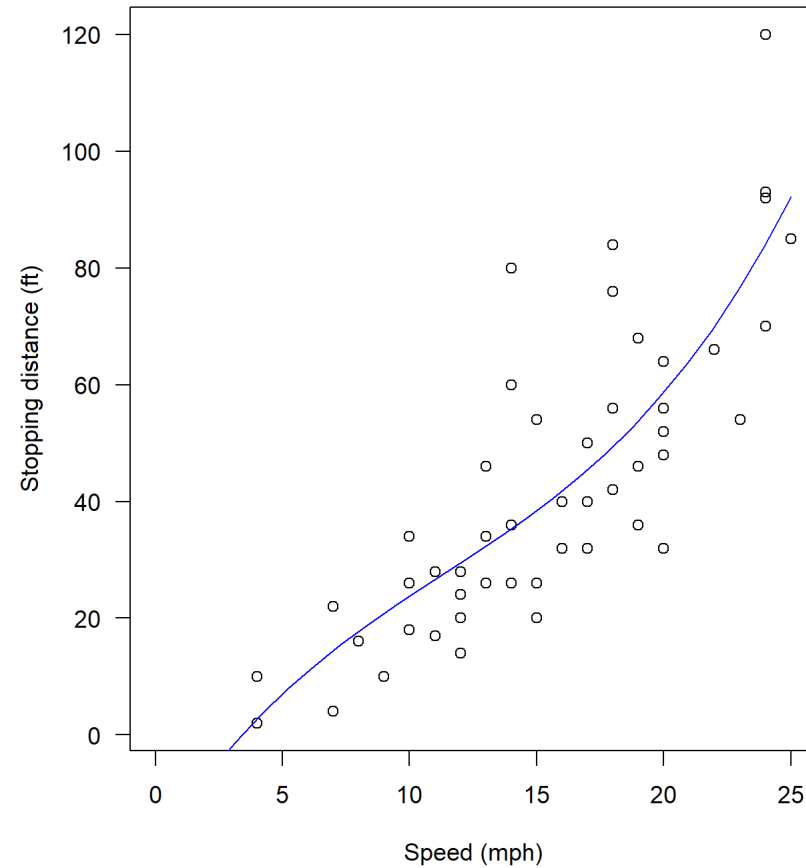
```
## An example of polynomial regression
plot(cars, xlab = "Speed (mph)",
     ylab = "Stopping distance (ft)",
     las = 1, xlim = c(0, 25))

lm(dist ~ poly(speed, 3),
   data = cars) ->
model

seq(0, 25, length.out = 25) ->
inputs_of_x

predict(model,
        data.frame(speed = inputs_of_x)) ->
prediction_y

lines(inputs_of_x,
      prediction_y,
      col = "blue")
```



And arithmetic operation

```
(4 + 5)
```

```
[1] 9
```

```
(4 + 5) /
```

```
6
```

```
[1] 1.5
```



```
(4 + 5) /  
6 *  
7
```

[1] 10.5

```
(4 + 5) /
```

```
6 *
```

```
7 -
```

```
3
```

```
[1] 7.5
```

```
(4 + 5) /  
6 *  
7 -  
3
```

```
1:10
```

```
[1] 7.5
```

```
[1] 1 2 3 4 5 6 7 8 9 10
```

```
(4 + 5) /  
6 *  
7 -  
3
```

```
1:10 %%  
3
```

```
[1] 7.5
```

```
[1] 1 2 0 1 2 0 1 2 0 1
```

```
(4 + 5) /  
6 *  
7 -  
3
```

```
1:10 %%  
3
```

```
1:10
```

```
[1] 7.5
```

```
[1] 1 2 0 1 2 0 1 2 0 1
```

```
[1] 1 2 3 4 5 6 7 8 9 10
```

```
(4 + 5) /  
6 *  
7 -  
3
```

```
1:10 %%  
3
```

```
1:10 %/%  
3
```

```
[1] 7.5
```

```
[1] 1 2 0 1 2 0 1 2 0 1
```

```
[1] 0 0 1 1 1 2 2 2 3 3
```

```
(4 + 5) /  
6 *  
7 -  
3
```

```
1:10 %%  
3
```

```
1:10 %/%  
3
```

```
33
```

```
[1] 7.5
```

```
[1] 1 2 0 1 2 0 1 2 0 1
```

```
[1] 0 0 1 1 1 2 2 2 3 3
```

```
[1] 33
```

```
(4 + 5) /  
6 *  
7 -  
3
```

```
1:10 %%  
3
```

```
1:10 %/%  
3
```

```
33 %%  
15
```

```
[1] 7.5
```

```
[1] 1 2 0 1 2 0 1 2 0 1
```

```
[1] 0 0 1 1 1 2 2 2 3 3
```

```
[1] 3
```



```
(4 + 5) /  
  6 *  
  7 -  
  3
```

```
1:10 %%  
  3
```

```
1:10 %/%  
  3
```

```
33 %%  
  15
```

```
4
```

```
[1] 7.5
```

```
[1] 1 2 0 1 2 0 1 2 0 1
```

```
[1] 0 0 1 1 1 2 2 2 3 3
```

```
[1] 3
```

```
[1] 4
```

```
(4 + 5) /  
  6 *  
  7 -  
  3
```

```
1:10 %%  
  3
```

```
1:10 %/%  
  3
```

```
33 %%  
  15
```

```
4 %/%  
  2
```

```
[1] 7.5
```

```
[1] 1 2 0 1 2 0 1 2 0 1
```

```
[1] 0 0 1 1 1 2 2 2 3 3
```

```
[1] 3
```

```
[1] 2
```

```
(4 + 5) /  
  6 *  
  7 -  
  3
```

```
1:10 %%  
  3
```

```
1:10 %/%  
  3
```

```
33 %%  
  15
```

```
4 %/%  
  2
```

```
4
```

```
[1] 7.5
```

```
[1] 1 2 0 1 2 0 1 2 0 1
```

```
[1] 0 0 1 1 1 2 2 2 3 3
```

```
[1] 3
```

```
[1] 2
```

```
[1] 4
```

```
(4 + 5) /
```

```
6 *
```

```
7 -
```

```
3
```

```
1:10 %%
```

```
3
```

```
1:10 %/%
```

```
3
```

```
33 %%
```

```
15
```

```
4 %/%
```

```
2
```

```
4 ^
```

```
5
```

```
[1] 7.5
```

```
[1] 1 2 0 1 2 0 1 2 0 1
```

```
[1] 0 0 1 1 1 2 2 2 3 3
```

```
[1] 3
```

```
[1] 2
```

```
[1] 1024
```

```

(4 + 5) /
  6 *
  7 -
  3

1:10 %%
  3

1:10 %/%
  3

33 %%
  15

4 %/%
  2

4 ^
  5

```

```
matrix(1:4, ncol = 1)
```

```
[1] 7.5
```

```
[1] 1 2 0 1 2 0 1 2 0 1
```

```
[1] 0 0 1 1 1 2 2 2 3 3
```

```
[1] 3
```

```
[1] 2
```

```
[1] 1024
```

```
[,1]
```

```
[1,] 1
```

```
[2,] 2
```

```
[3,] 3
```

```
[4,] 4
```

```
(4 + 5) /  
  6 *  
  7 -  
  3
```

```
1:10 %%  
  3
```

```
1:10 %/%  
  3
```

```
33 %%  
  15
```

```
4 %/%  
  2
```

```
4 ^  
  5
```

```
matrix(1:4, ncol = 1) %*%  
  matrix(1:4, nrow = 1)
```

```
[1] 7.5
```

```
[1] 1 2 0 1 2 0 1 2 0 1
```

```
[1] 0 0 1 1 1 2 2 2 3 3
```

```
[1] 3
```

```
[1] 2
```

```
[1] 1024
```

	[,1]	[,2]	[,3]	[,4]
[1,]	1	2	3	4
[2,]	2	4	6	8
[3,]	3	6	9	12
[4,]	4	8	12	16

```

(4 + 5) /
  6 *
  7 -
  3

1:10 %%
  3

1:10 %/%
  3

33 %%
  15

4 %/%
  2

4 ^
  5

matrix(1:4, ncol = 1) %*%
  matrix(1:4, nrow = 1)

matrix(1:4, ncol = 4)

```

```

[1] 7.5

[1] 1 2 0 1 2 0 1 2 0 1

[1] 0 0 1 1 1 2 2 2 3 3

[1] 3

[1] 2

[1] 1024

      [,1] [,2] [,3] [,4]
[1,]    1    2    3    4
[2,]    2    4    6    8
[3,]    3    6    9   12
[4,]    4    8   12   16

      [,1] [,2] [,3] [,4]
[1,]    1    2    3    4

```

```

(4 + 5) /
  6 *
  7 -
  3

1:10 %%
  3

1:10 %/%
  3

33 %%
  15

4 %/%
  2

4 ^
  5

matrix(1:4, ncol = 1) %*%
  matrix(1:4, nrow = 1)

matrix(1:4, ncol = 4) %*%
  matrix(1:4, nrow = 4)

```

```

[1] 7.5

[1] 1 2 0 1 2 0 1 2 0 1

[1] 0 0 1 1 1 2 2 2 3 3

[1] 3

[1] 2

[1] 1024

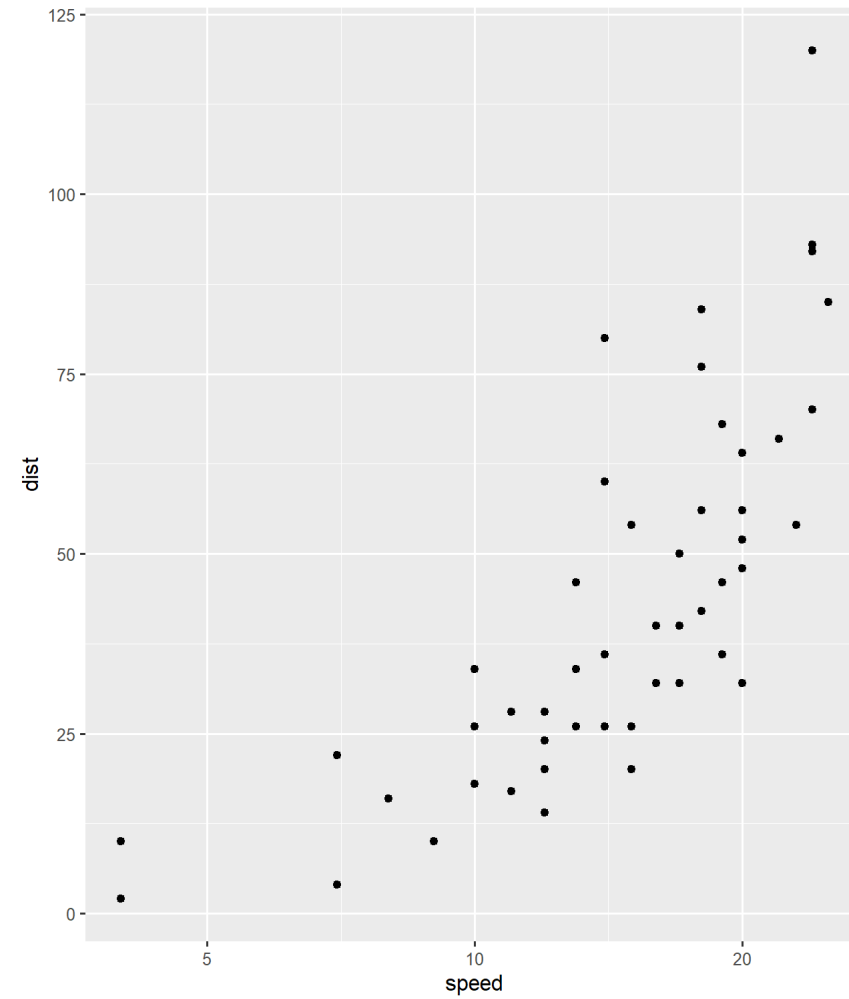
      [,1] [,2] [,3] [,4]
[1,]    1    2    3    4
[2,]    2    4    6    8
[3,]    3    6    9   12
[4,]    4    8   12   16

      [,1]
[1,]    30

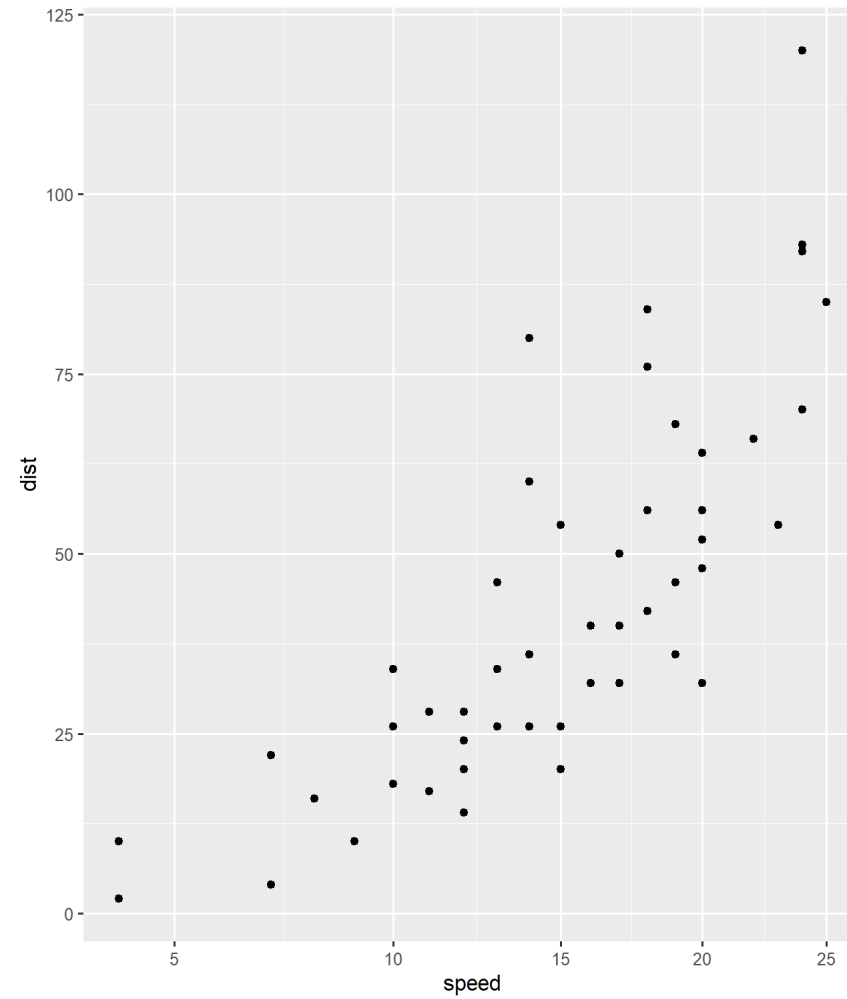
```



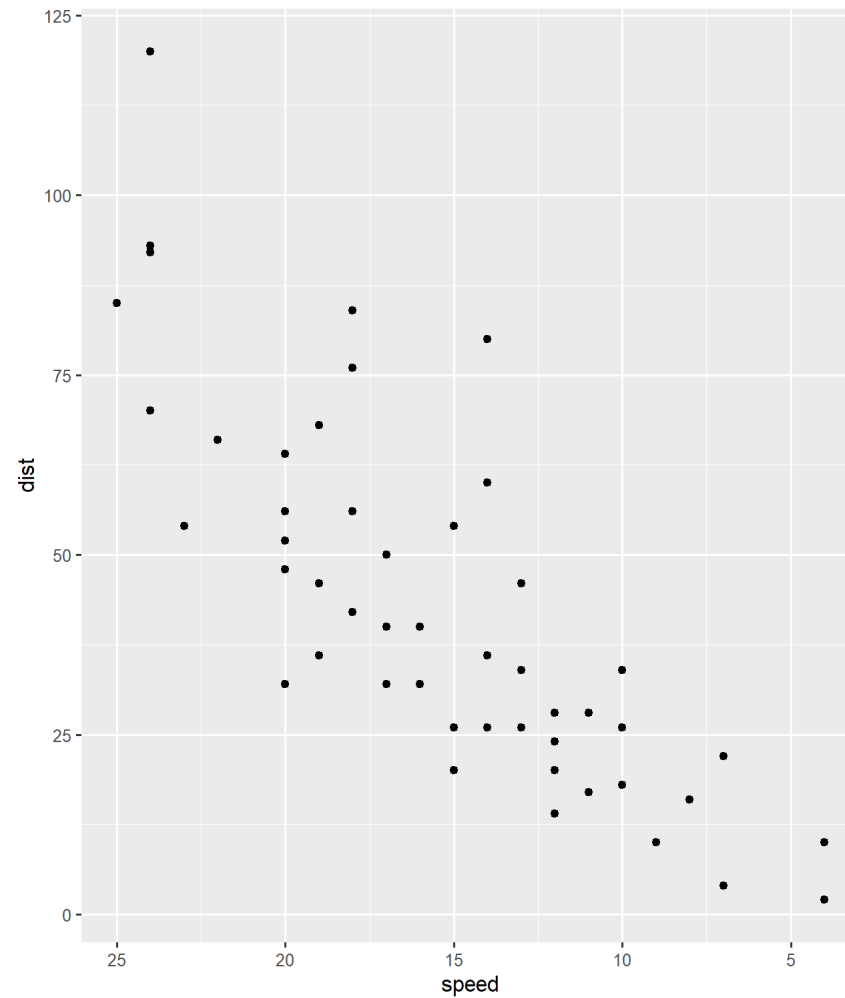
```
ggplot(data = cars) +  
  aes(x = speed) +  
  aes(y = dist) +  
  geom_point() +  
  scale_x_log10()
```



```
ggplot(data = cars) +  
  aes(x = speed) +  
  aes(y = dist) +  
  geom_point() +  
  scale_x_sqrt()
```



```
ggplot(data = cars) +  
  aes(x = speed) +  
  aes(y = dist) +  
  geom_point() +  
  scale_x_reverse()
```



A new addition is the `%%$%` pipe from the `magrittr` library. And example follows.

```
library(magrittr)
```

```
library(magrittr)
```

```
cars
```

	speed	dist
1	4	2
2	4	10
3	7	4
4	7	22
5	8	16
6	9	10
7	10	18
8	10	26
9	10	34
10	11	17
11	11	28
12	12	14
13	12	20
14	12	24
15	12	28
16	13	26
17	13	34
18	13	34
19	13	46
20	14	26
21	14	36
22	14	60
23	14	80
24	15	20
25	15	26
26	15	54
27	16	32
28	16	40
29	17	32
30	17	40
31	17	50
32	18	42

```
library(magrittr)
```

```
cars %$%
```

```
  cor(x = speed,  
       y = dist)
```

```
[1] 0.8068949
```

Custom Styling

Pipe to correlation coefficient

```
library(magrittr)
```

Pipe to correlation coefficient

```
library(magrittr)
```

```
cars
```

	speed	dist
1	4	2
2	4	10
3	7	4
4	7	22
5	8	16
6	9	10
7	10	18
8	10	26
9	10	34
10	11	17
11	11	28
12	12	14
13	12	20
14	12	24
15	12	28
16	13	26
17	13	34
18	13	34
19	13	46
20	14	26
21	14	36
22	14	60
23	14	80
24	15	20
25	15	26

Pipe to correlation coefficient

```
library(magrittr)
```

```
cars %$%
```

```
  cor(x = speed,  
      y = dist)
```

```
[1] 0.8068949
```

hello

```
library(magrittr)
```

goodbye

```
library(magrittr)
```

```
cars
```

$$\frac{\sum(1)}{2}$$

```
library(magrittr)
```

```
cars %$%
```

```
  cor(x = speed,  
       y = dist)
```

hello

goodbye

	speed	dist
1	4	2
2	4	10
3	7	4
4	7	22
5	8	16
6	9	10
7	10	18
8	10	26
9	10	34
10	11	17
11	11	28
12	12	14
13	12	20
14	12	24
15	12	28
16	13	26
17	13	34
18	13	34
19	13	46
20	14	26
21	14	36
22	14	60
23	14	80
24	15	20
25	15	26
26	15	54
27	16	32
28	16	40
29	17	32
30	17	40
31	17	50
32	18	42

[1] 0.8068949

$$\frac{\Sigma(1)}{2}$$

hello

goodbye

$$\frac{\Sigma(1)}{2}$$

```
cars
```

	speed	dist
1	4	2
2	4	10
3	7	4
4	7	22
5	8	16
6	9	10
7	10	18
8	10	26
9	10	34
10	11	17
11	11	28
12	12	14
13	12	20
14	12	24
15	12	28
16	13	26
17	13	34
18	13	34
19	13	46
20	14	26
21	14	36
22	14	60
23	14	80
24	15	20
25	15	26
26	15	54
27	16	32
28	16	40
29	17	32
30	17	40
31	17	50
32	18	42

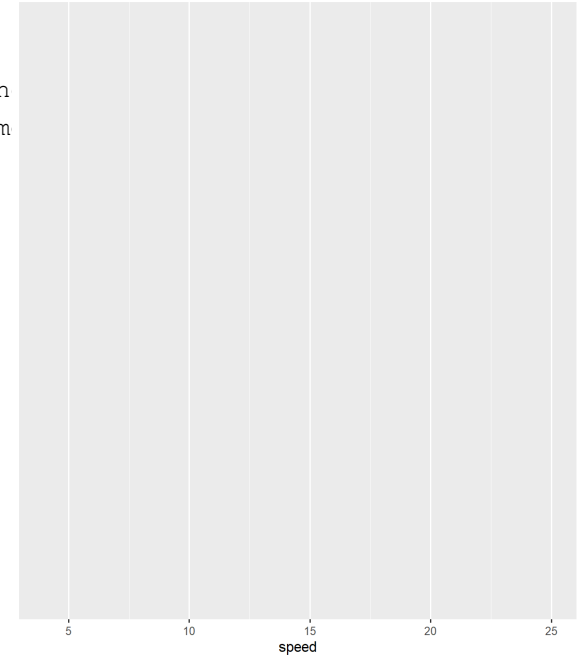
```
cars %>%
```

```
  ggplot()
```

```
function (data = NULL, mapping = aes(), ...  
{  
  UseMethod("ggplot")  
}  
<bytecode: 0x00000226c24fe700>  
<environment: namespace:ggplot2>
```

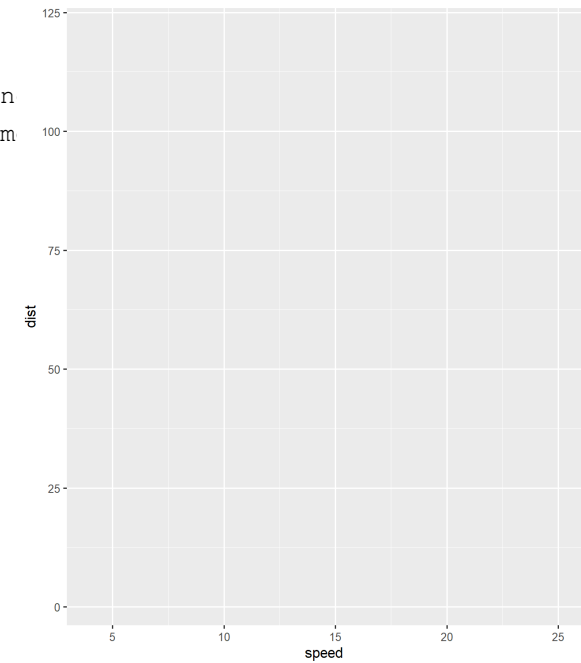
```
cars %>%  
  ggplot() +  
  aes(x = speed)
```

```
function (x, y, ...)  
{  
  exprs <- enquos(x = x, y = y, ..., .ign  
  aes <- new_aes(exprs, env = parent.frame  
  rename_aes(aes)  
}  
<bytecode: 0x00000226c0cede20>  
<environment: namespace:ggplot2>
```



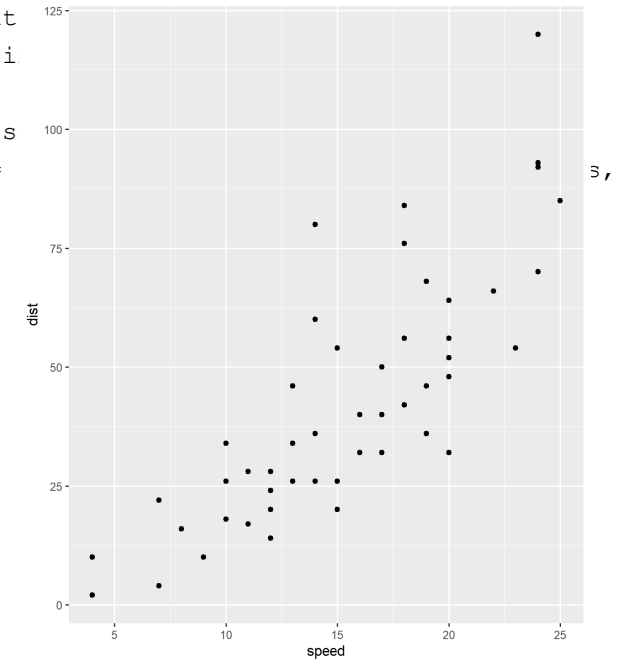
```
cars %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist)
```

```
function (x, y, ...)  
{  
  exprs <- enquos(x = x, y = y, ..., .ign  
  aes <- new_aes(exprs, env = parent.frame  
  rename_aes(aes)  
}  
<bytecode: 0x00000226c0cede20>  
<environment: namespace:ggplot2>
```



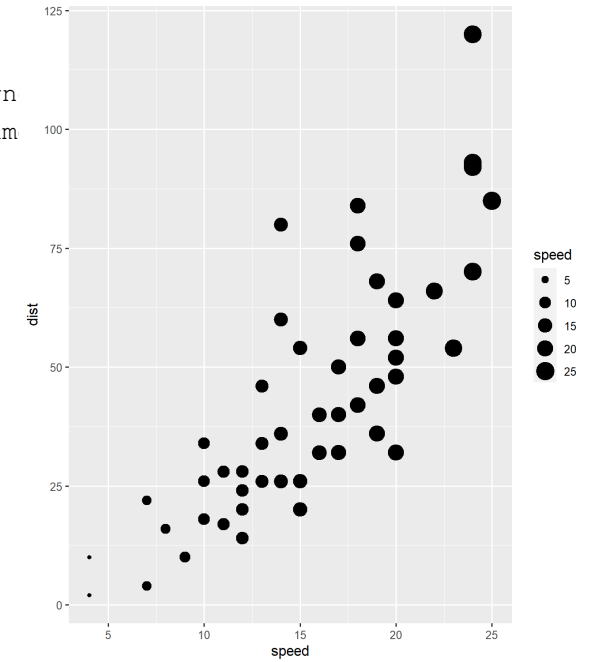

```
cars %>%  
  ggplot() +  
  aes(x = speed) +  
  aes(y = dist) +  
  geom_point()
```

```
function (mapping = NULL, data = NULL, stat  
  ..., na.rm = FALSE, show.legend = NA, i  
{  
  layer(data = data, mapping = mapping, s  
    position = position, show.legend =  
    params = list(na.rm = na.rm, ...))  
}  
<bytecode: 0x00000226c2357670>  
<environment: namespace:ggplot2>
```



```
cars %>%
  ggplot() +
    aes(x = speed) +
    aes(y = dist) +
    geom_point() +
    aes(size = speed)
```

```
function (x, y, ...)
{
  exprs <- enquos(x = x, y = y, ..., .ign
  aes <- new_aes(exprs, env = parent.frame
  rename_aes(aes)
}
<bytecode: 0x00000226c0cede20>
<environment: namespace:ggplot2>
```



Slow Message

This

is

my

text.

'One driver of equality we should invest in is upskilling everyone - not just the select few.'

digital inclusion.

Xaringan slide show look and feel

To quickly change the look and feel of your {xaringan} slide show, you might check out the [available themes](#) from the xaringan package and [xaringanthemer](#) package.

Another extremely useful resource for xaringan styling is Alison Hill's "[Meet xaringan: Making slides in R Markdown](#)".

Sharing your flipbooks

Flipbooks created with Xaringan are multi-file creations. The figures produced are stored separately from the main html document. This presents a little bit of a challenge for sharing your work. You can zip up all the associated files and share that way. Alternatively, you can share as a website. I've shared my work on github with github pages.

- flip, zip, and ship
- get it on github, with github pages. A good walk through is the one that I learned with (to get [the ggplot2 flipbook](#) online) by Brian Caffo <https://www.youtube.com/watch?v=BBCesiebEuQ> Larger flipbooks will take longer to load online - something to keep in mind as you are building.

The flipbooked portion of this presentation was created with the new `{flipbookr}` package. Get it with
`remotes::install_github("EvaMaeRey/flipbookr")`