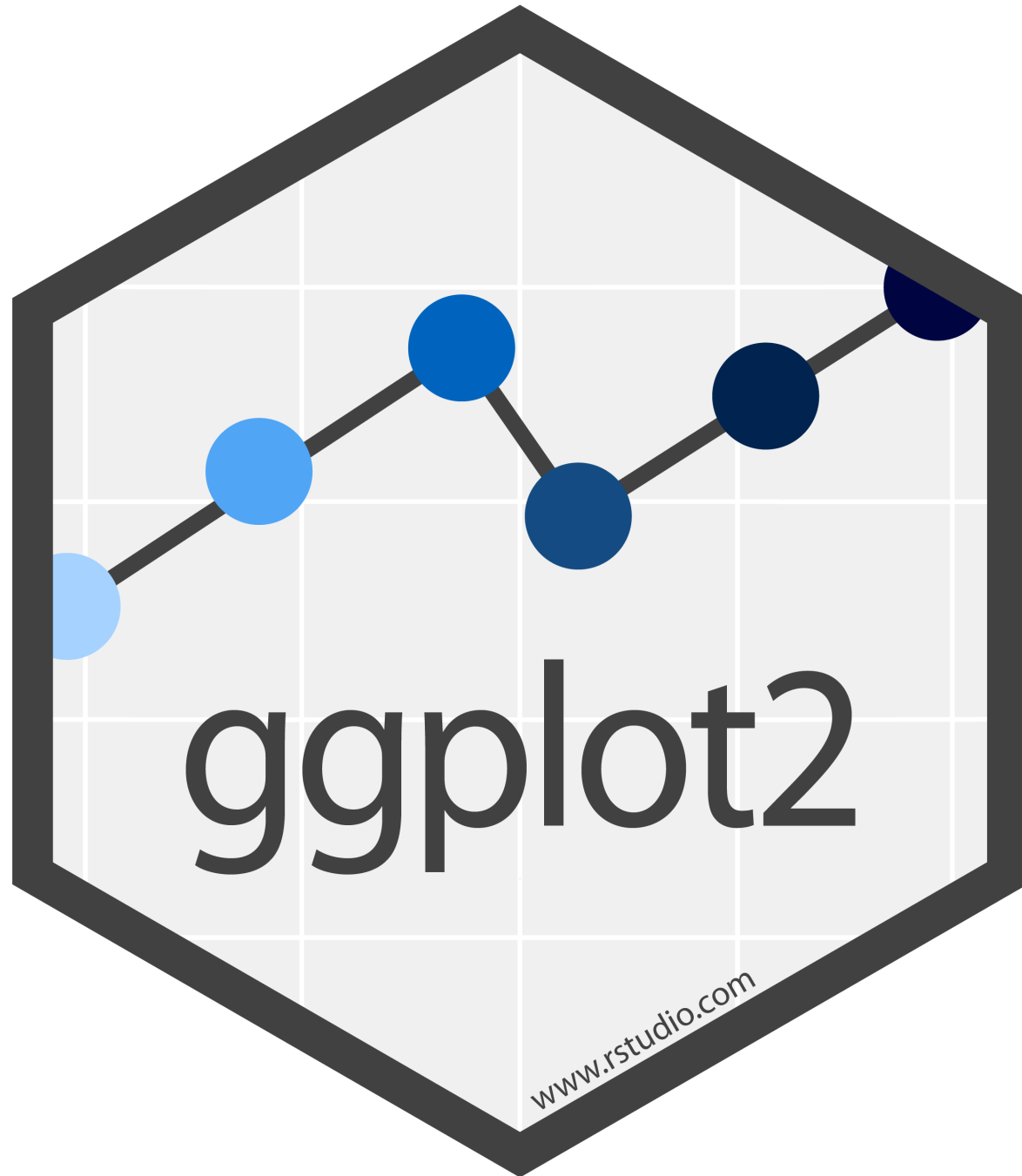


# Visualization with ggplot2

Lecture 10

Dr. Colin Rundel



# The Grammar of Graphics

- Conceptualized by Leland Wilkinson in The Grammar of Graphics (1999)
- Attempt to taxonomize the basic elements of statistical graphics
- Adapted for R by Hadley Wickham (2009)
  - consistent and compact syntax to describe statistical graphics
  - highly modular - breaks down graphs into semantic components
  - not meant as a guide on which graph to use or how to best convey your data (more on that next time), but it does have some strong opinions.

# Terminology

A statistical graphic is a...

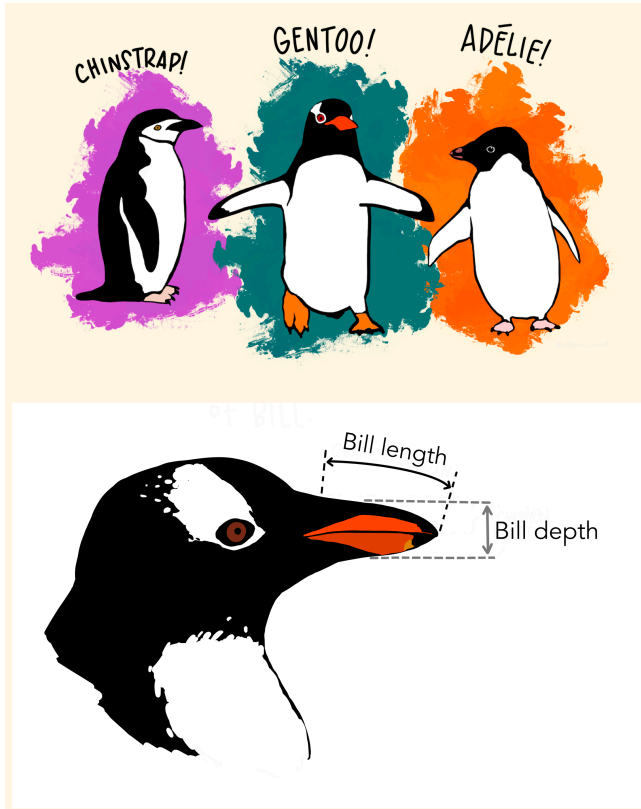
- mapping of **data**
- which may be **statistically transformed** (summarized, log-transformed, etc.)
- to **aesthetic attributes** (color, size, xy-position, etc.)
- using **geometric objects** (points, lines, bars, etc.)
- and mapped onto a specific **facet** and **coordinate system**

# Anatomy of a ggplot call

```
1 ggplot(  
2   data = [dataframe],  
3   mapping = aes(  
4     x = [var x], y = [var y],  
5     color = [var color],  
6     shape = [var shape],  
7     ...  
8   )  
9 ) +  
10 geom_[some geom](  
11   mapping = aes(  
12     color = [var geom color],  
13     ...  
14   )  
15 ) +  
16 ... # other geometries  
17 scale_[some axis]_[some scale]() +  
18 facet_[some facet]([formula]) +  
19 ... # other options
```

# Data - Palmer Penguins

Measurements of penguin species, island in the Palmer Archipelago, size (flipper length, body mass, bill dimensions), and sex.



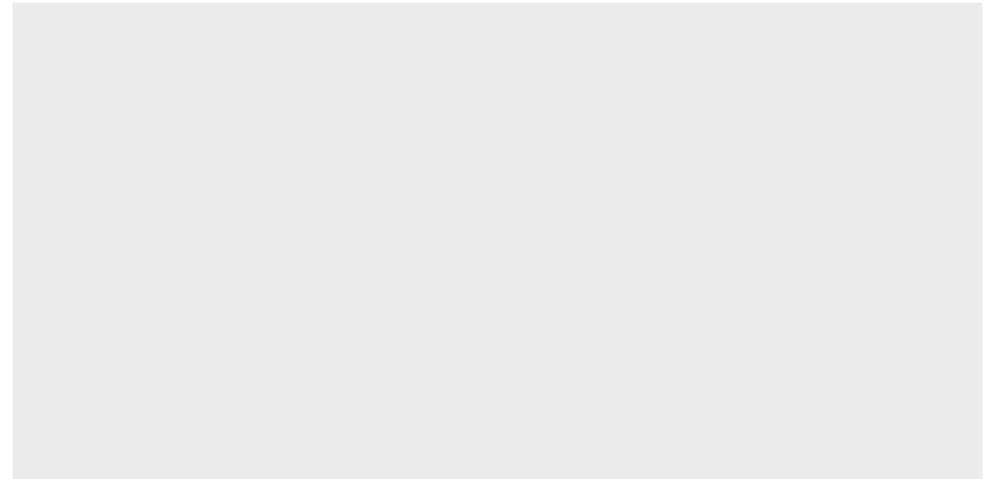
```
1 library(palmerpenguins)
2 penguins
```

```
# A tibble: 344 × 8
  species island bill_length_mm bill_depth_mm
  <fct>   <fct>         <dbl>         <dbl>
1 Adelie  Torgersen      39.1           18.7
2 Adelie  Torgersen      39.5           17.4
3 Adelie  Torgersen      40.3           18
4 Adelie  Torgersen      NA             NA
5 Adelie  Torgersen      36.7           19.3
6 Adelie  Torgersen      39.3           20.6
7 Adelie  Torgersen      38.9           17.8
8 Adelie  Torgersen      39.2           19.6
9 Adelie  Torgersen      34.1           18.1
10 Adelie Torgersen      42             20.2
# i 334 more rows
# i 4 more variables: flipper_length_mm <int>,
#   body_mass_g <int>, sex <fct>, year <int>
```

# Text $\leftrightarrow$ Plot

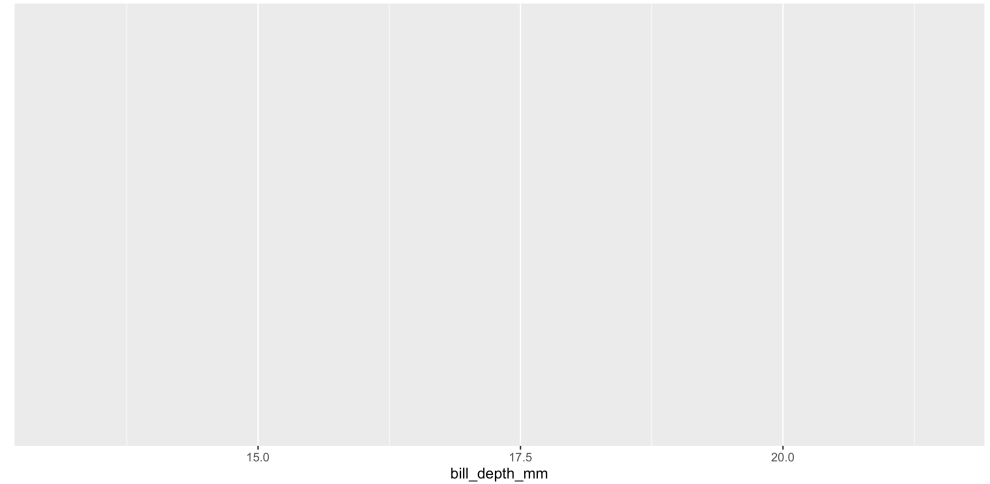
Start with the *penguins* data frame

```
1 ggplot(data = penguins)
```



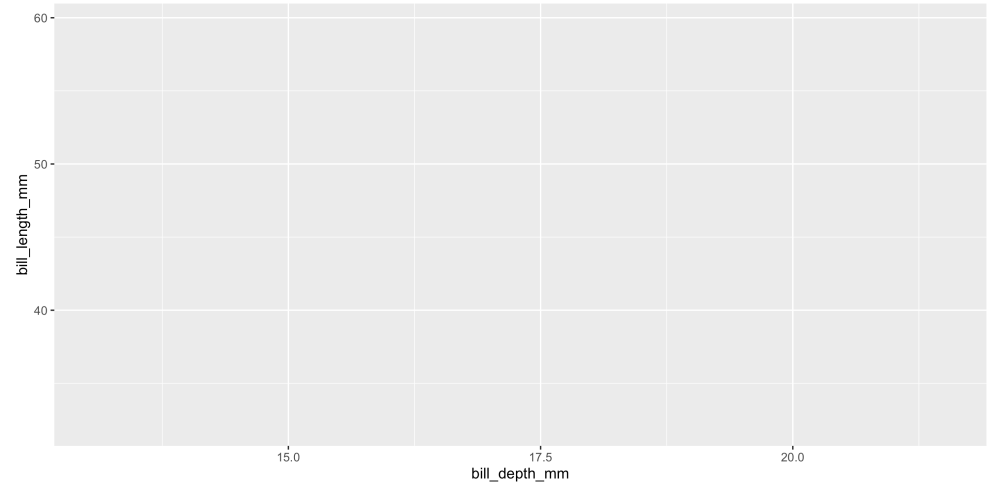
Start with the `penguins` data frame, *map bill depth to the x-axis*

```
1 ggplot(  
2   data = penguins,  
3   mapping = aes(  
4     x = bill_depth_mm  
5   )  
6 )
```



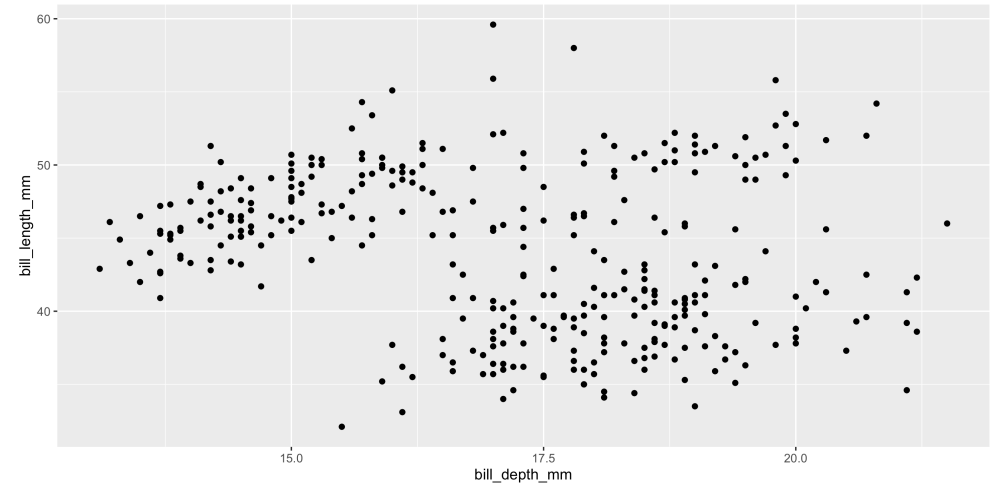
Start with the `penguins` data frame, map bill depth to the x-axis *and* map bill length to the y-axis.

```
1 ggplot(  
2   data = penguins,  
3   mapping = aes(  
4     x = bill_depth_mm,  
5     y = bill_length_mm  
6   )  
7 )
```



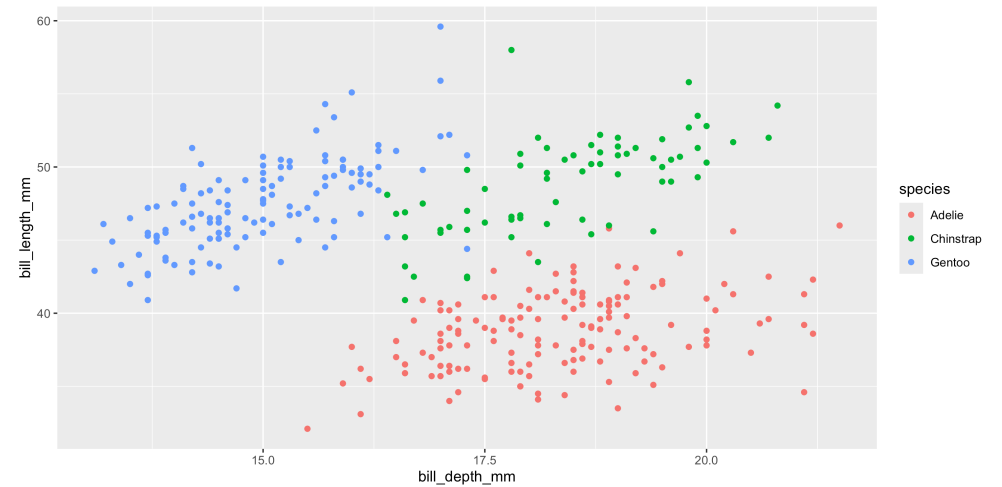
Start with the `penguins` data frame, map bill depth to the x-axis and map bill length to the y-axis. *Represent each observation with a point*

```
1 ggplot(  
2   data = penguins,  
3   mapping = aes(  
4     x = bill_depth_mm,  
5     y = bill_length_mm  
6   )  
7 ) +  
8   geom_point()
```



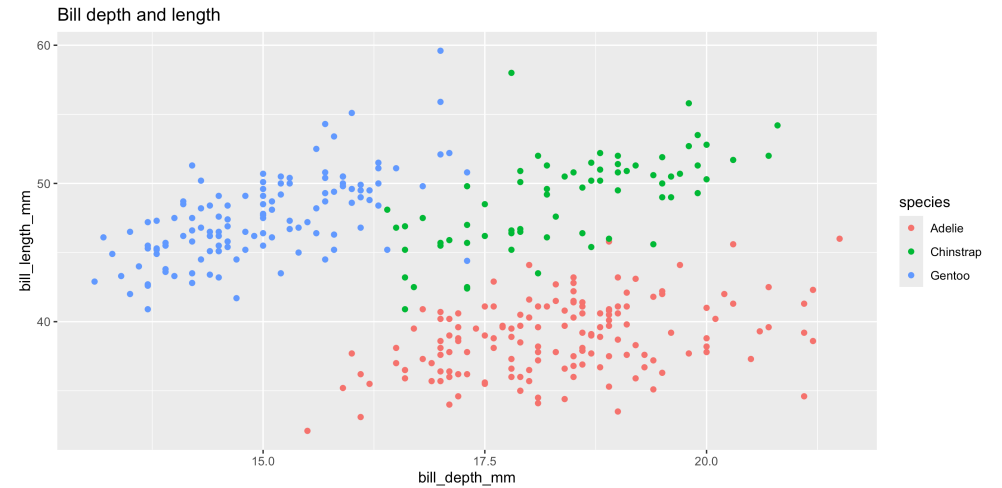
Start with the `penguins` data frame, map bill depth to the x-axis and map bill length to the y-axis. Represent each observation with a point *and map species to the color of each point*.

```
1 ggplot(  
2   data = penguins,  
3   mapping = aes(  
4     x = bill_depth_mm,  
5     y = bill_length_mm  
6   )  
7 ) +  
8 geom_point(  
9   mapping = aes(color = species)  
10 )
```



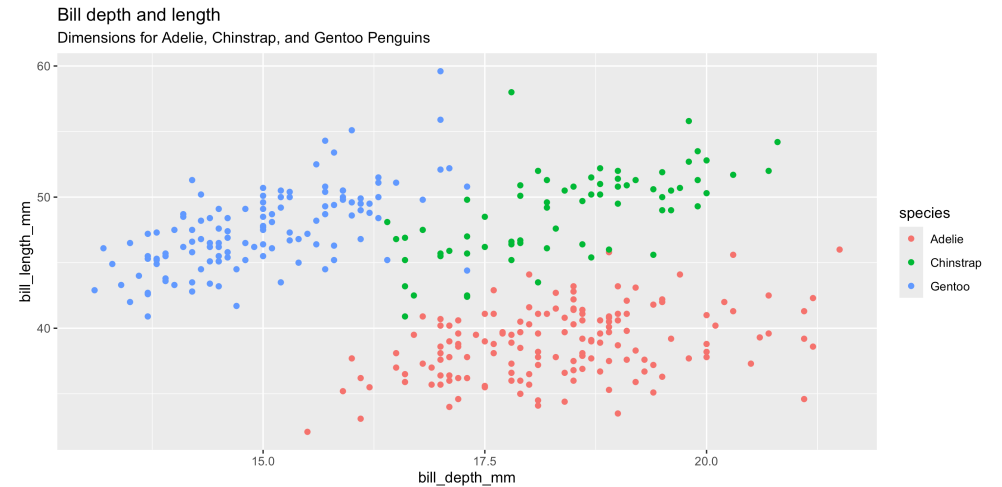
Start with the `penguins` data frame, map bill depth to the x-axis and map bill length to the y-axis. Represent each observation with a point and map species to the color of each point. *Title the plot "Bill depth and length"*

```
1 ggplot(  
2   data = penguins,  
3   mapping = aes(  
4     x = bill_depth_mm,  
5     y = bill_length_mm  
6   )  
7 ) +  
8 geom_point(  
9   mapping = aes(color = species)  
10 ) +  
11 labs(title = "Bill depth and length")
```



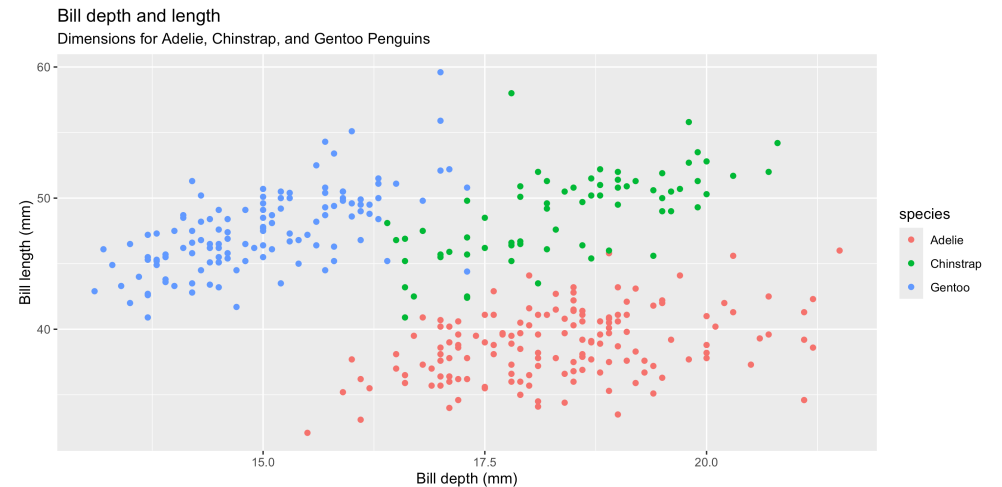
Start with the `penguins` data frame, map bill depth to the x-axis and map bill length to the y-axis. Represent each observation with a point and map species to the color of each point. Title the plot “Bill depth and length”, *add the subtitle “Dimensions for Adelie, Chinstrap, and Gentoo Penguins”*

```
1 ggplot(  
2   data = penguins,  
3   mapping = aes(  
4     x = bill_depth_mm,  
5     y = bill_length_mm  
6   )  
7 ) +  
8 geom_point(  
9   mapping = aes(color = species)  
10 ) +  
11 labs(  
12   title = "Bill depth and length",  
13   subtitle = paste(  
14     "Dimensions for Adelie,",  
15     "Chinstrap, and Gentoo",  
16     "Penguins")  
17 )
```



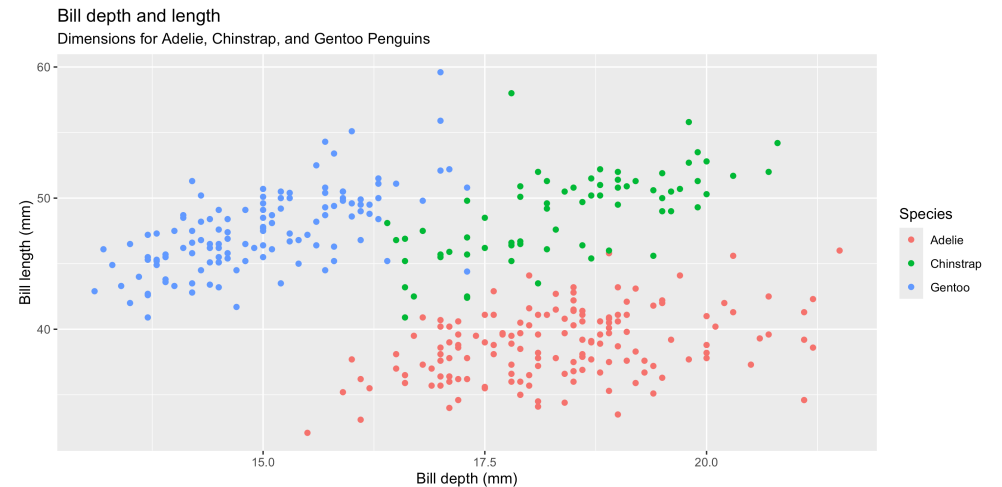
Start with the `penguins` data frame, map bill depth to the x-axis and map bill length to the y-axis. Represent each observation with a point and map species to the color of each point. Title the plot “Bill depth and length”, add the subtitle “Dimensions for Adelie, Chinstrap, and Gentoo Penguins”, *label the x and y axes as “Bill depth (mm)” and “Bill length (mm)”*, respectively

```
1 ggplot(  
2   data = penguins,  
3   mapping = aes(  
4     x = bill_depth_mm,  
5     y = bill_length_mm  
6   )  
7 ) +  
8 geom_point(  
9   mapping = aes(color = species)  
10 ) +  
11 labs(  
12   title = "Bill depth and length",  
13   subtitle = paste(  
14     "Dimensions for Adelie,",  
15     "Chinstrap, and Gentoo",  
16     "Penguins"),  
17   x = "Bill depth (mm)",  
18   y = "Bill length (mm)"  
19 )
```



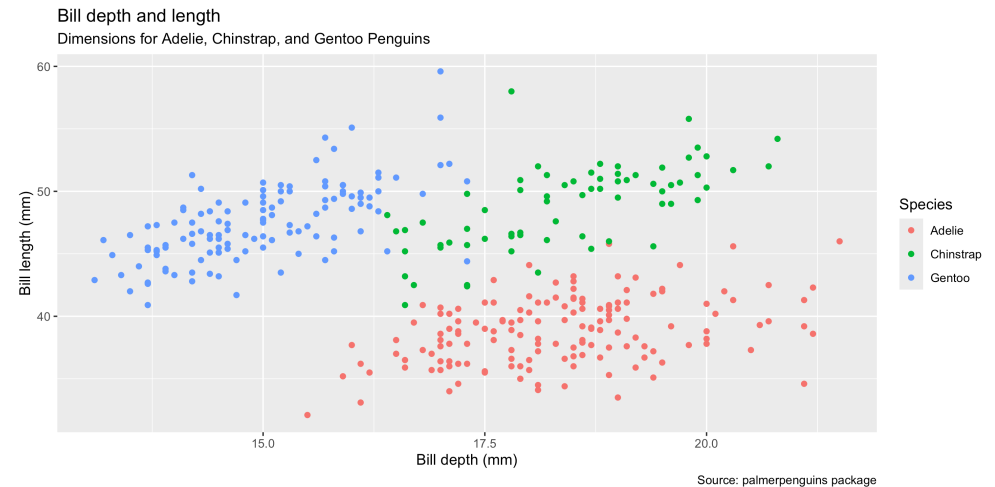
Start with the `penguins` data frame, map bill depth to the x-axis and map bill length to the y-axis. Represent each observation with a point and map species to the color of each point. Title the plot “Bill depth and length”, add the subtitle “Dimensions for Adelie, Chinstrap, and Gentoo Penguins”, label the x and y axes as “Bill depth (mm)” and “Bill length (mm)”, respectively, *label the legend “Species”*

```
1 ggplot(  
2   data = penguins,  
3   mapping = aes(  
4     x = bill_depth_mm,  
5     y = bill_length_mm  
6   )  
7 ) +  
8 geom_point(  
9   mapping = aes(color = species)  
10 ) +  
11 labs(  
12   title = "Bill depth and length",  
13   subtitle = paste(  
14     "Dimensions for Adelie,",  
15     "Chinstrap, and Gentoo",  
16     "Penguins"),  
17   x = "Bill depth (mm)",  
18   y = "Bill length (mm)",  
19   color = "Species"  
20 )
```



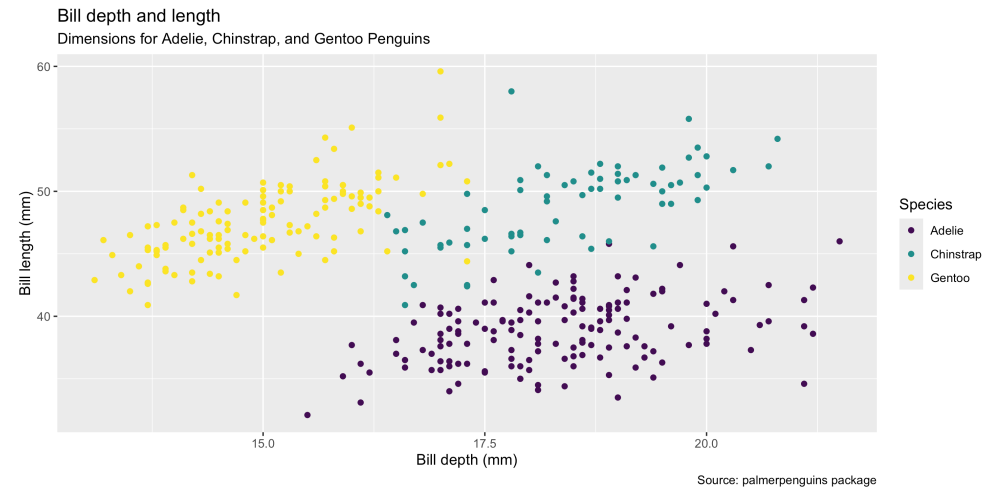
Start with the `penguins` data frame, map bill depth to the x-axis and map bill length to the y-axis. Represent each observation with a point and map species to the color of each point. Title the plot “Bill depth and length”, add the subtitle “Dimensions for Adelie, Chinstrap, and Gentoo Penguins”, label the x and y axes as “Bill depth (mm)” and “Bill length (mm)”, respectively, label the legend “Species”, and add a caption for the data source.

```
1 ggplot(  
2   data = penguins,  
3   mapping = aes(  
4     x = bill_depth_mm,  
5     y = bill_length_mm  
6   )  
7 ) +  
8 geom_point(  
9   mapping = aes(color = species)  
10 ) +  
11 labs(  
12   title = "Bill depth and length",  
13   subtitle = paste(  
14     "Dimensions for Adelie,",  
15     "Chinstrap, and Gentoo",  
16     "Penguins"),  
17   x = "Bill depth (mm)",  
18   y = "Bill length (mm)",  
19   color = "Species",  
20   caption = "Source: palmerpenguins package",  
21 )
```



Start with the `penguins` data frame, map bill depth to the x-axis and map bill length to the y-axis. Represent each observation with a point and map species to the color of each point. Title the plot “Bill depth and length”, add the subtitle “Dimensions for Adelie, Chinstrap, and Gentoo Penguins”, label the x and y axes as “Bill depth (mm)” and “Bill length (mm)”, respectively, label the legend “Species”, and add a caption for the data source. *Finally, use the viridis color palette for all points.*

```
1 ggplot(  
2   data = penguins,  
3   mapping = aes(  
4     x = bill_depth_mm,  
5     y = bill_length_mm  
6   )  
7 ) +  
8 geom_point(  
9   mapping = aes(color = species)  
10 ) +  
11 labs(  
12   title = "Bill depth and length",  
13   subtitle = paste(  
14     "Dimensions for Adelie,",  
15     "Chinstrap, and Gentoo",  
16     "Penguins"),  
17   x = "Bill depth (mm)",  
18   y = "Bill length (mm)",  
19   color = "Species",  
20   caption = "Source: palmerpenguins package"  
21 ) +  
22 scale_color_viridis_d()
```



# Aesthetics

# Aesthetics options

Commonly used characteristics of plotting geometries that can be **mapped to a specific variable** in the data, examples include:

- `x, y` (position)
- `color`
- `fill`
- `shape`
- `size`
- `alpha` (transparency)
- `linetype`

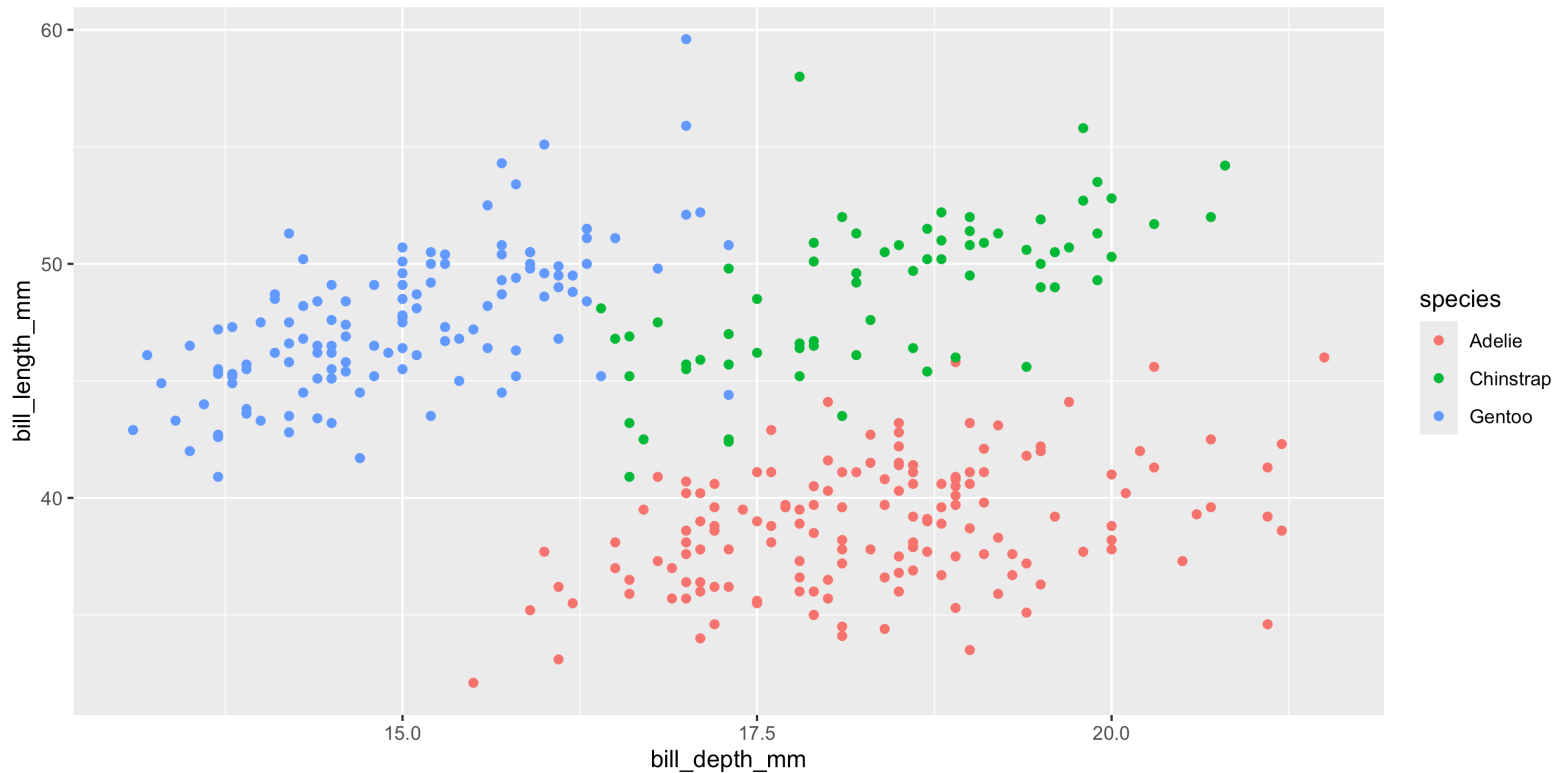
Different geometries have different aesthetics available - see the [ggplot2 geoms help files](#) for listings.

- Aesthetics given in `ggplot()` apply to all `geoms`.
- Aesthetics for a specific `geom_*()` can be overridden via `mapping` or as an argument.

# color

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(  
5     aes(color = species)  
6   )
```

Warning: Removed 2 rows containing missing values or values outside the scale range (``geom_point()``).

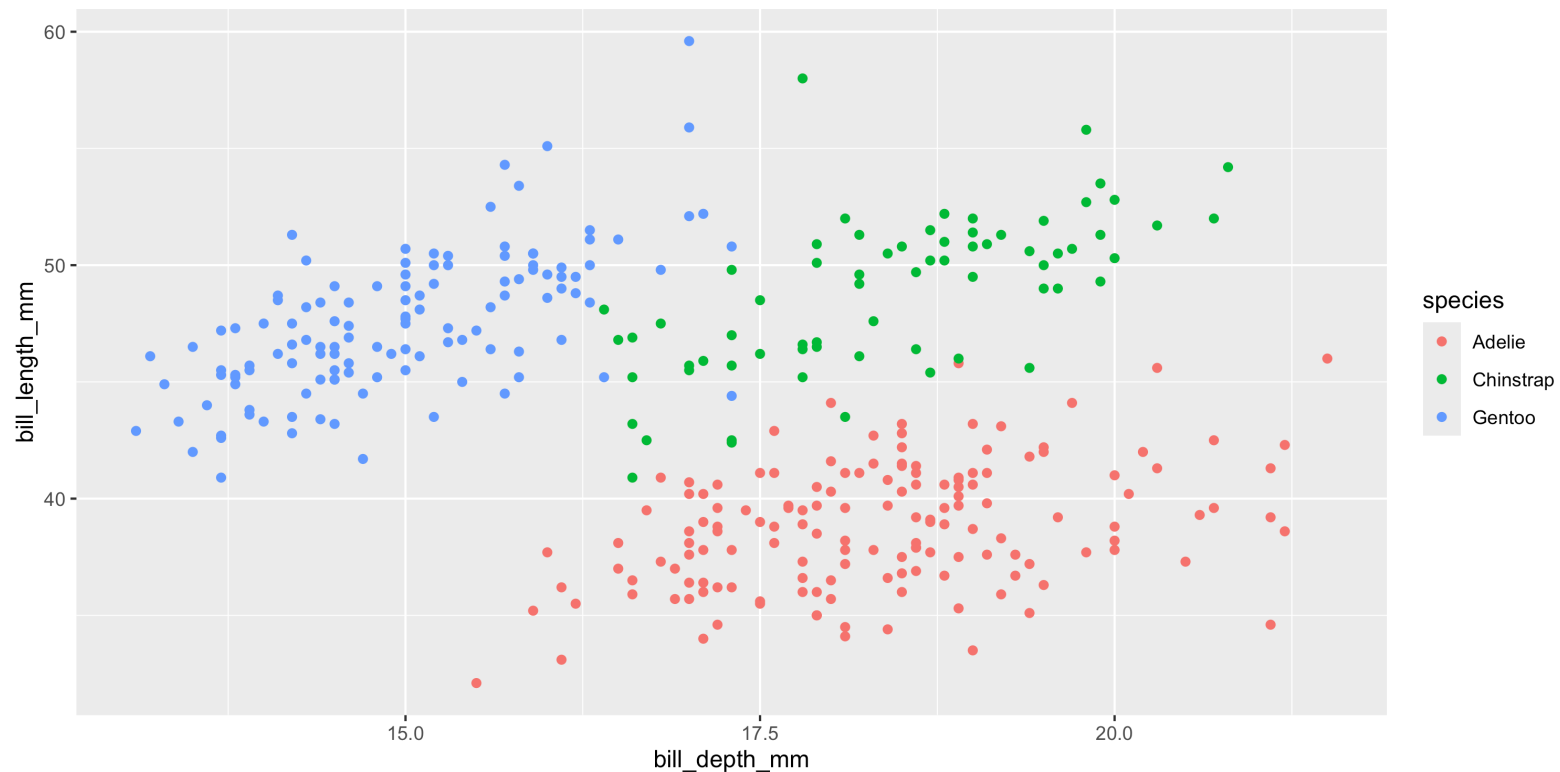


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# Avoid the warning

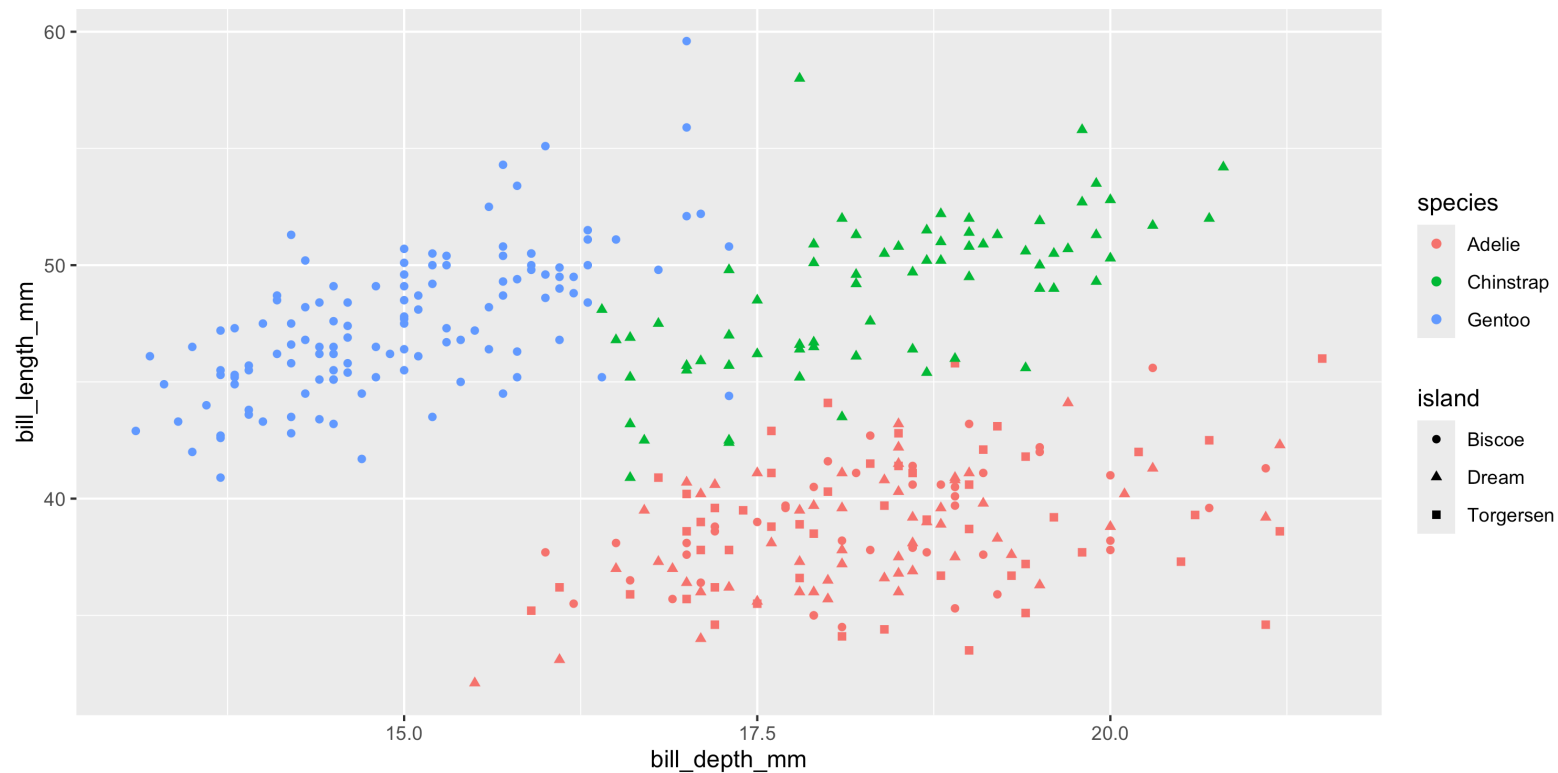
```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(  
5     aes(color = species), na.rm=TRUE  
6   )
```



# Shape

Mapped to a different variable than `color`

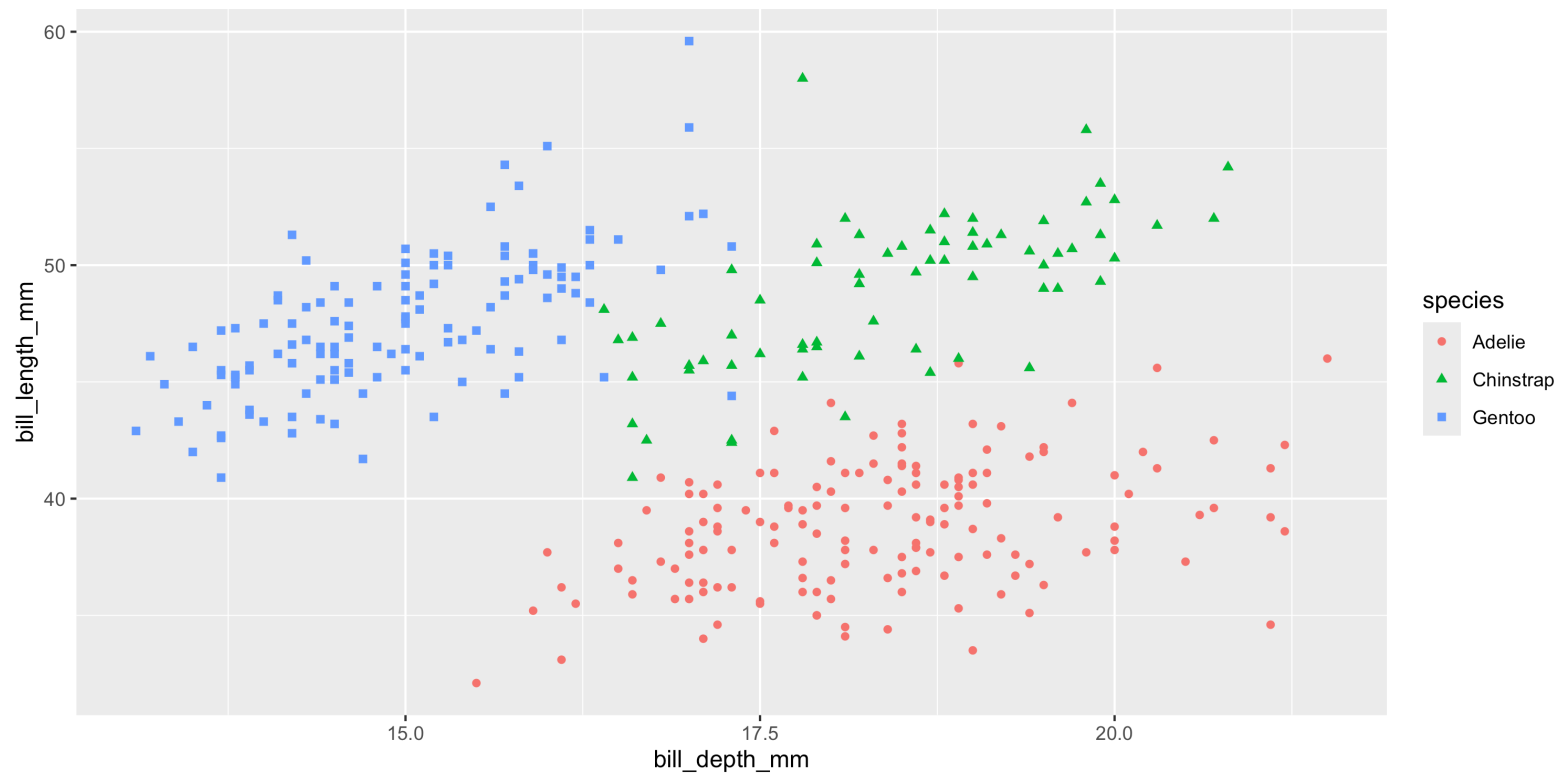
```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(  
5     aes(color = species, shape = island), na.rm = TRUE  
6   )
```



# Shape

Mapped to same variable as `color`

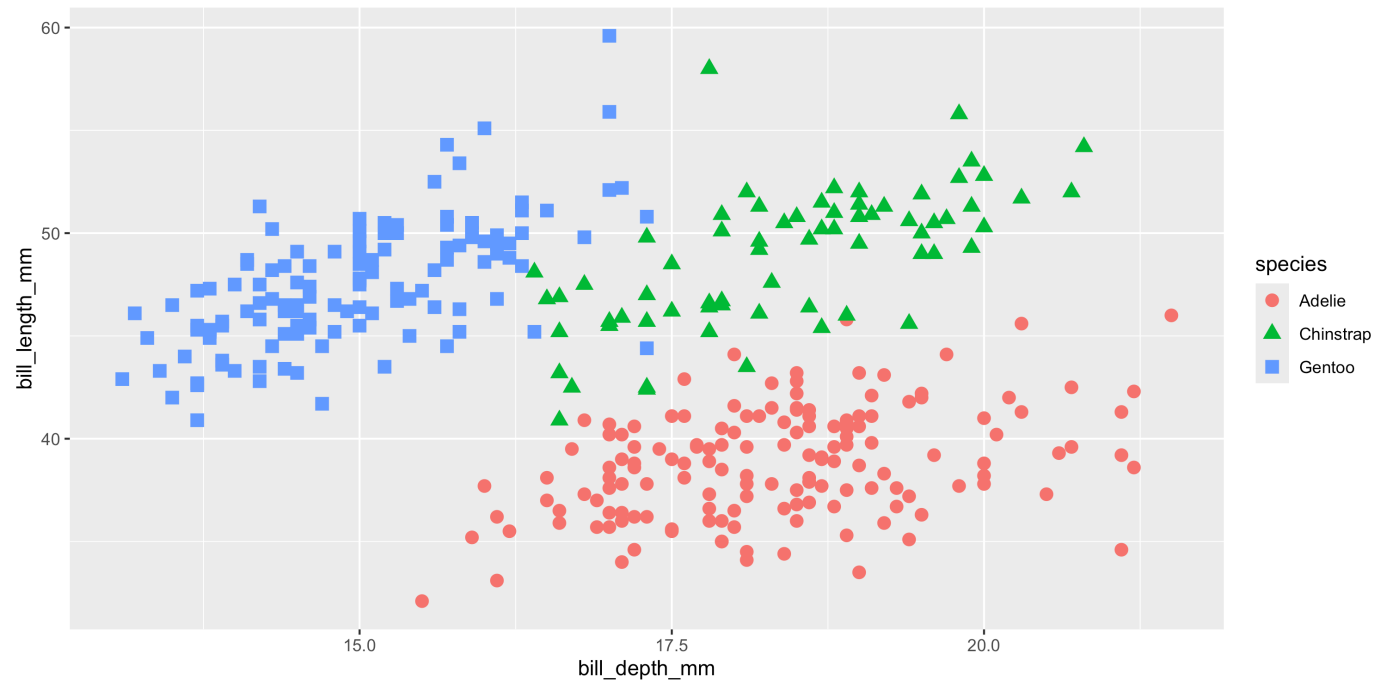
```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(  
5     aes(color = species, shape = species), na.rm = TRUE  
6   )
```



# Size

Using a fixed value - note that this value is outside of the `aes` call

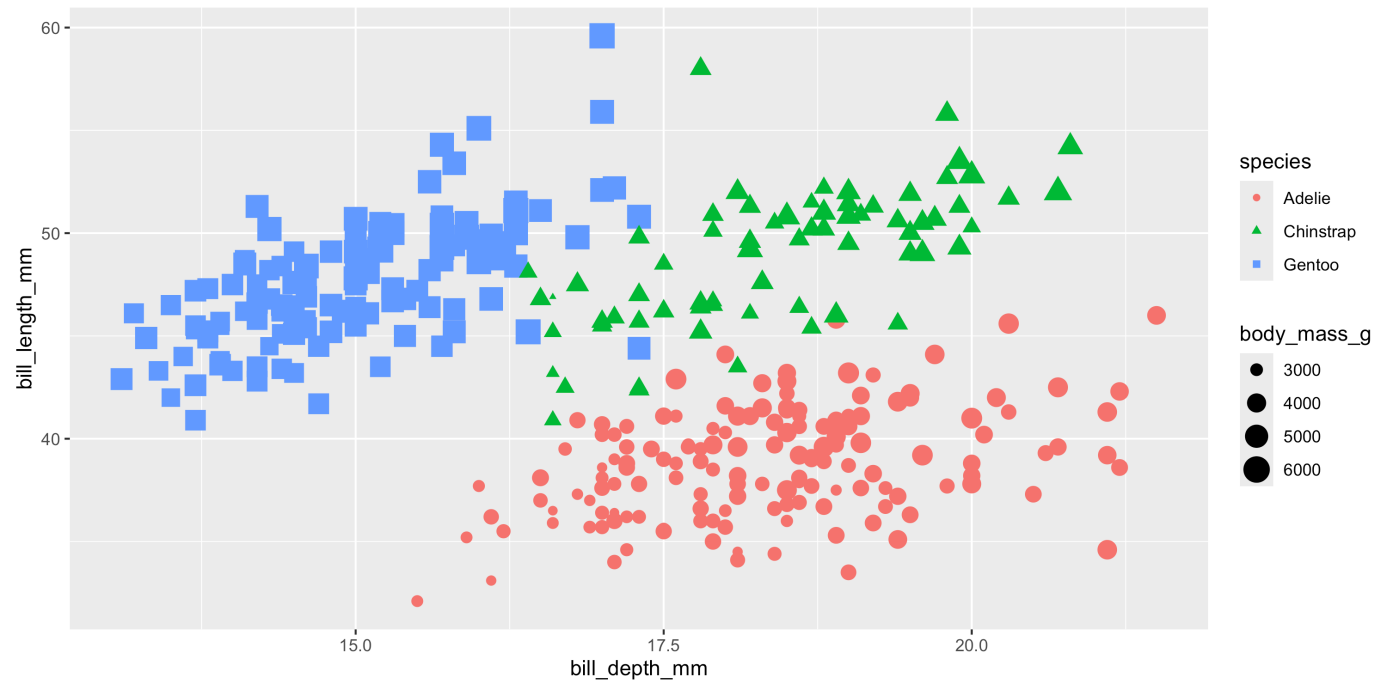
```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(  
5     aes(color = species, shape = species), na.rm = TRUE,  
6     size = 3  
7   )
```



# Size

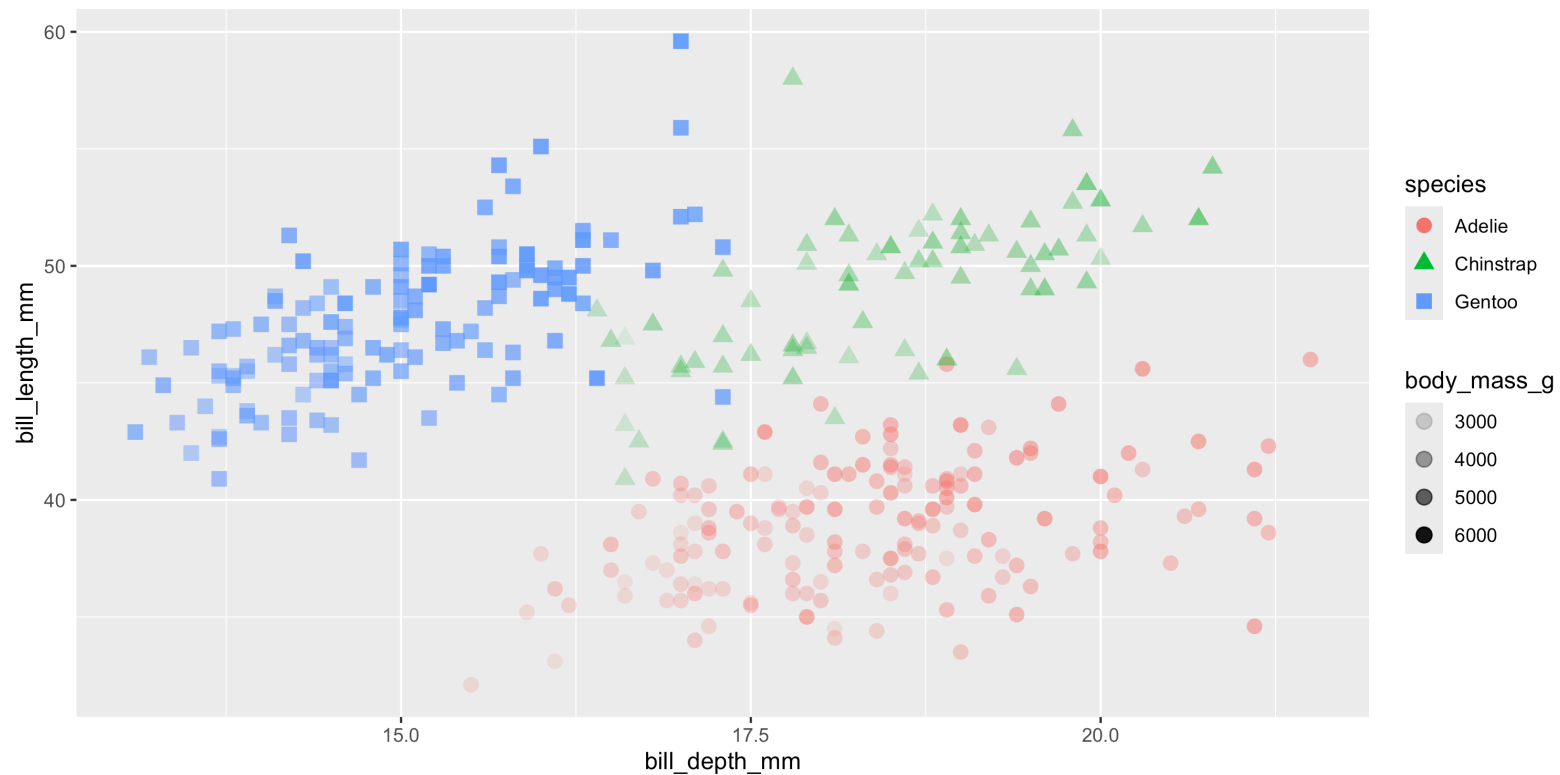
Mapped to a variable

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(  
5     aes(color = species, shape = species, size = body_mass_g), na.rm = TRUE  
6   )
```



# Alpha

```
1 ggplot(  
2   penguins,  
3   aes(x = bill_depth_mm, y = bill_length_mm)  
4 ) +  
5   geom_point(  
6     aes(color = species, shape = species, alpha = body_mass_g), na.rm = TRUE,  
7     size = 3  
8   )
```



# Mapping vs settings

- **Mapping** - Determine an aesthetic (the size, alpha, etc.) of a geom based on the values of a variable in the data
  - wrapped by `aes()` and pass as `mapping` argument to `ggplot()` or `geom_*()`.
- **Setting** - Determine an aesthetic (the size, alpha, etc.) of a geom using a constant value not directly from the data.
  - passed directly into `geom_*()` as an argument.

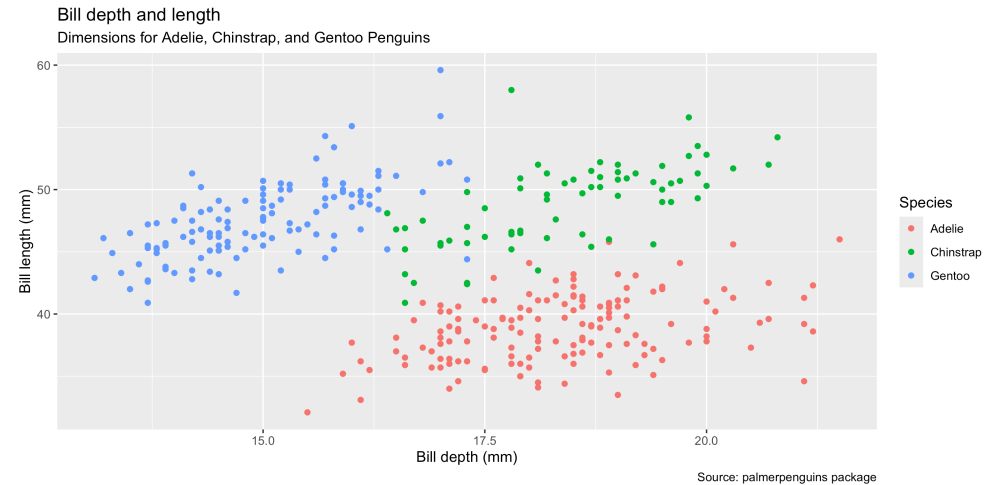
In this example `color`, `shape`, and `alpha` are all **mapping** while `size` is a **setting**.

# Labels

# labs()

In our previous example we saw the use of `labs()` to provide human readable labels to various plot elements.

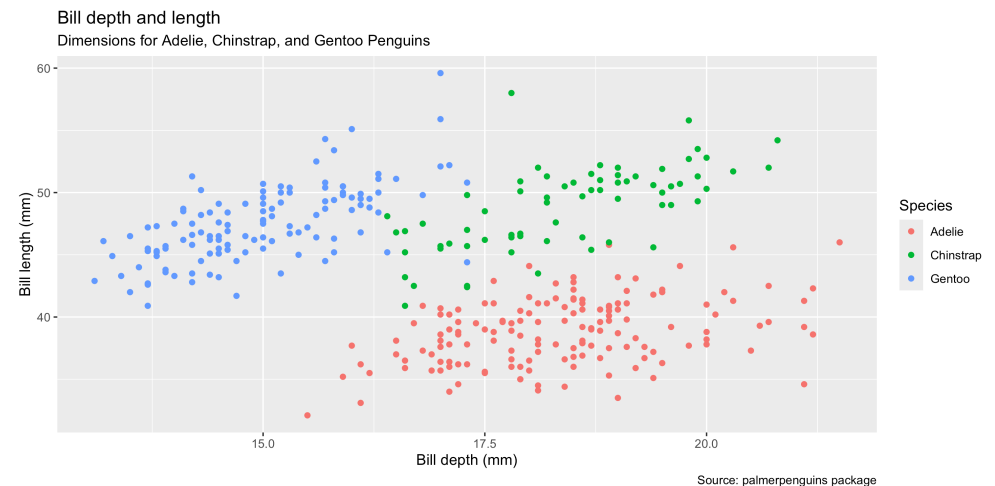
```
1 ggplot(  
2   data = penguins,  
3   mapping = aes(  
4     x = bill_depth_mm,  
5     y = bill_length_mm  
6   )  
7 ) +  
8 geom_point(  
9   mapping = aes(color = species), na.rm = TRUE  
10 ) +  
11 labs(  
12   title = "Bill depth and length",  
13   subtitle = paste(  
14     "Dimensions for Adelie,",  
15     "Chinstrap, and Gentoo",  
16     "Penguins"),  
17   x = "Bill depth (mm)",  
18   y = "Bill length (mm)",  
19   color = "Species",  
20   caption = "Source: palmerpenguins package"  
21 )
```



# Labels

Instead of overriding with `labs()`, we can annotate the data so that the label is generated automatically, by attaching a `label` attribute to the appropriate column in our data frame.

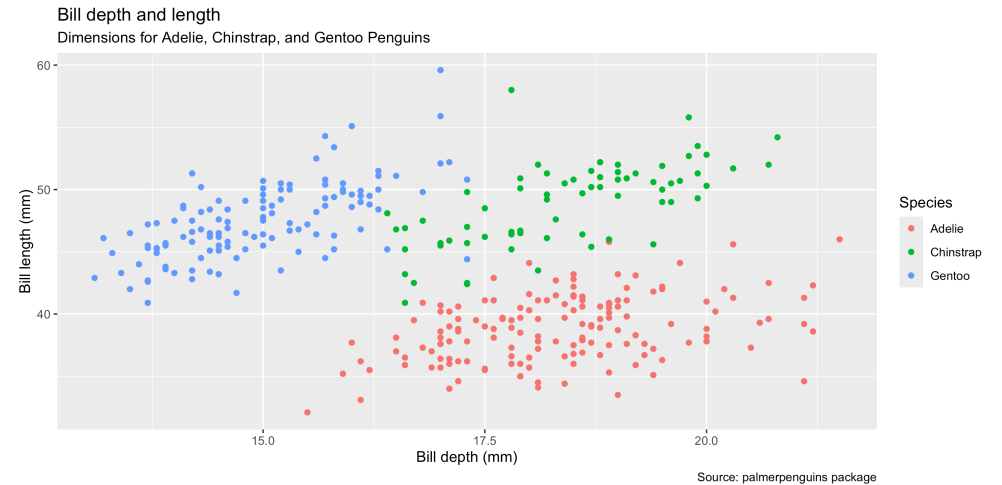
```
1 p_labeled = penguins
2 attr(p_labeled$species, "label") = "Species"
3 attr(p_labeled$bill_depth_mm, "label") = "Bill depth (mm)"
4 attr(p_labeled$bill_length_mm, "label") = "Bill length (mm)"
5
6 ggplot(
7   data = p_labeled,
8   mapping = aes(
9     x = bill_depth_mm,
10    y = bill_length_mm
11  )
12 ) +
13 geom_point(
14   mapping = aes(color = species), na.rm = TRUE
15 ) +
16 labs(
17   title = "Bill depth and length",
18   subtitle = paste(
19     "Dimensions for Adelie,",
20     "Chinstrap, and Gentoo",
21     "Penguins"),
22   caption = "Source: palmerpenguins package"
23 )
```



# Dictionary

Alternatively, we can provide a dictionary / lookup table via the `dictionary` argument of `labs()`.

```
1 lookup = c(
2   species = "Species",
3   bill_depth_mm = "Bill depth (mm)",
4   bill_length_mm = "Bill length (mm)"
5 )
6
7 ggplot(
8   data = penguins,
9   mapping = aes(
10    x = bill_depth_mm,
11    y = bill_length_mm
12  )
13 ) +
14 geom_point(
15   mapping = aes(color = species), na.rm = TRUE
16 ) +
17 labs(
18   title = "Bill depth and length",
19   subtitle = paste(
20     "Dimensions for Adelie,",
21     "Chinstrap, and Gentoo",
22     "Penguins"),
23   caption = "Source: palmerpenguins package",
24   dictionary = lookup
25 )
```

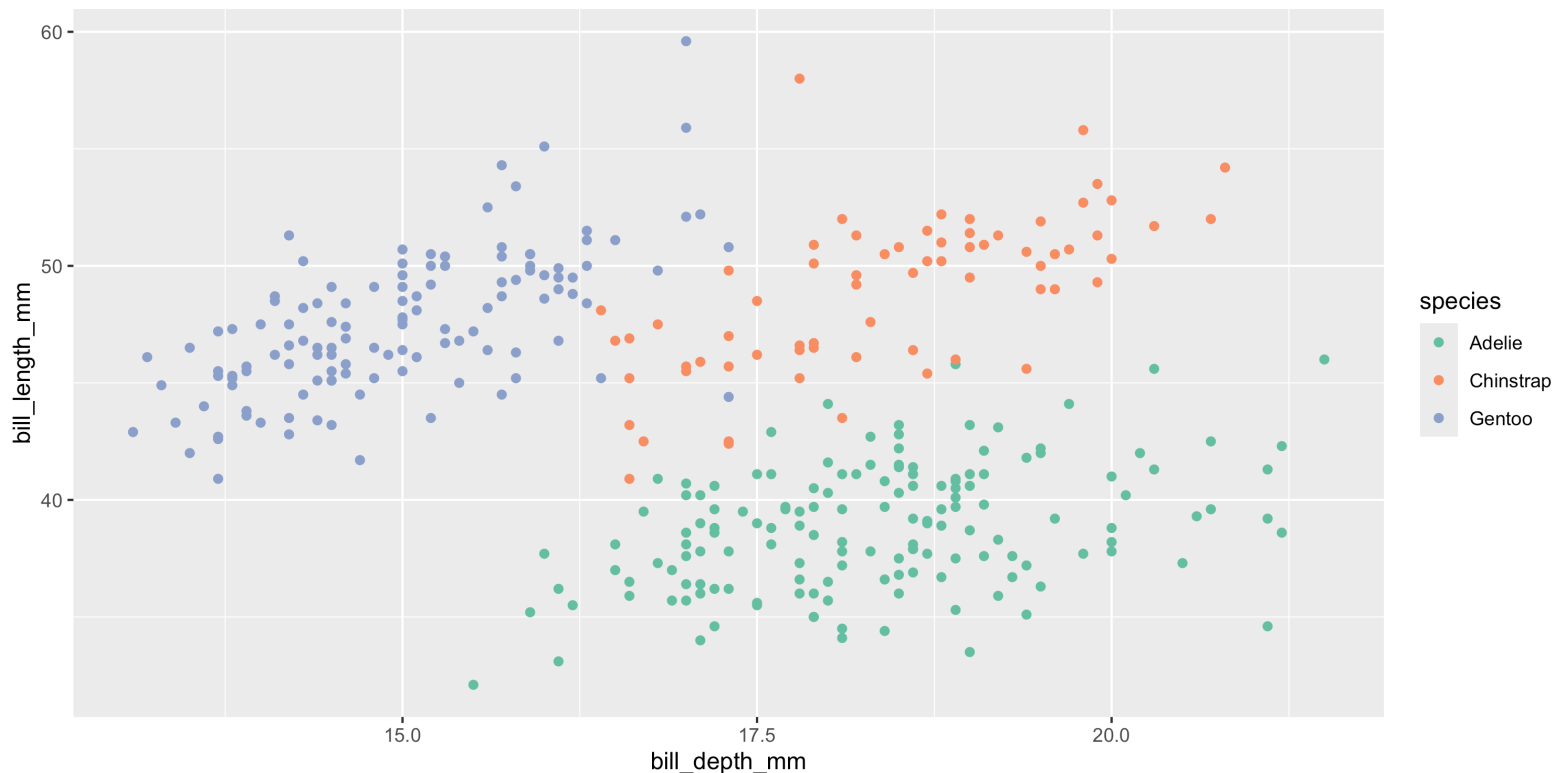


# Scales

# Scales

Scales control the mapping from data values to aesthetic values — they determine *how* a variable is translated into `color`, `size`, position, etc. Every aesthetic has a default scale, but we can override it:

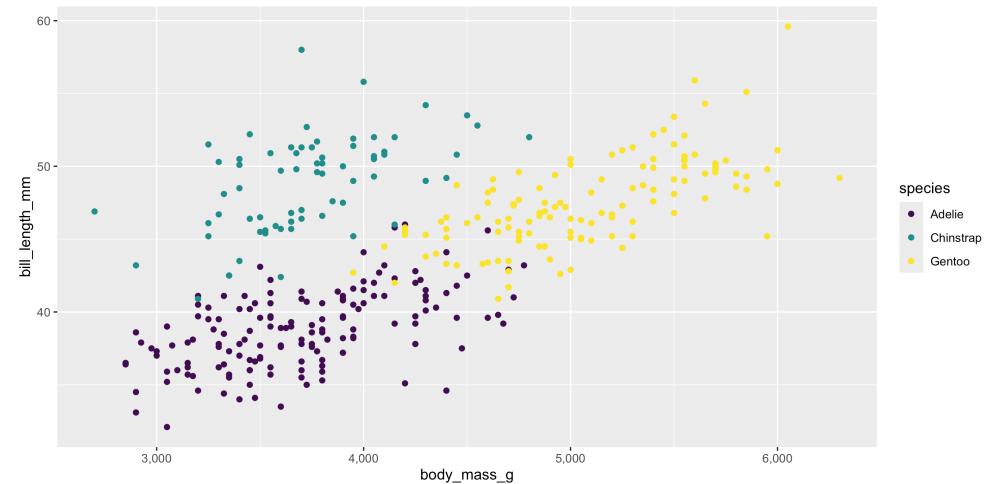
```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(aes(color = species), na.rm = TRUE) +  
5   scale_color_brewer(type = "qual", palette = "Set2")
```



# Axis scales

Scales also apply to positional aesthetics (`x` and `y`) — useful for transforming axes or customizing breaks and labels.

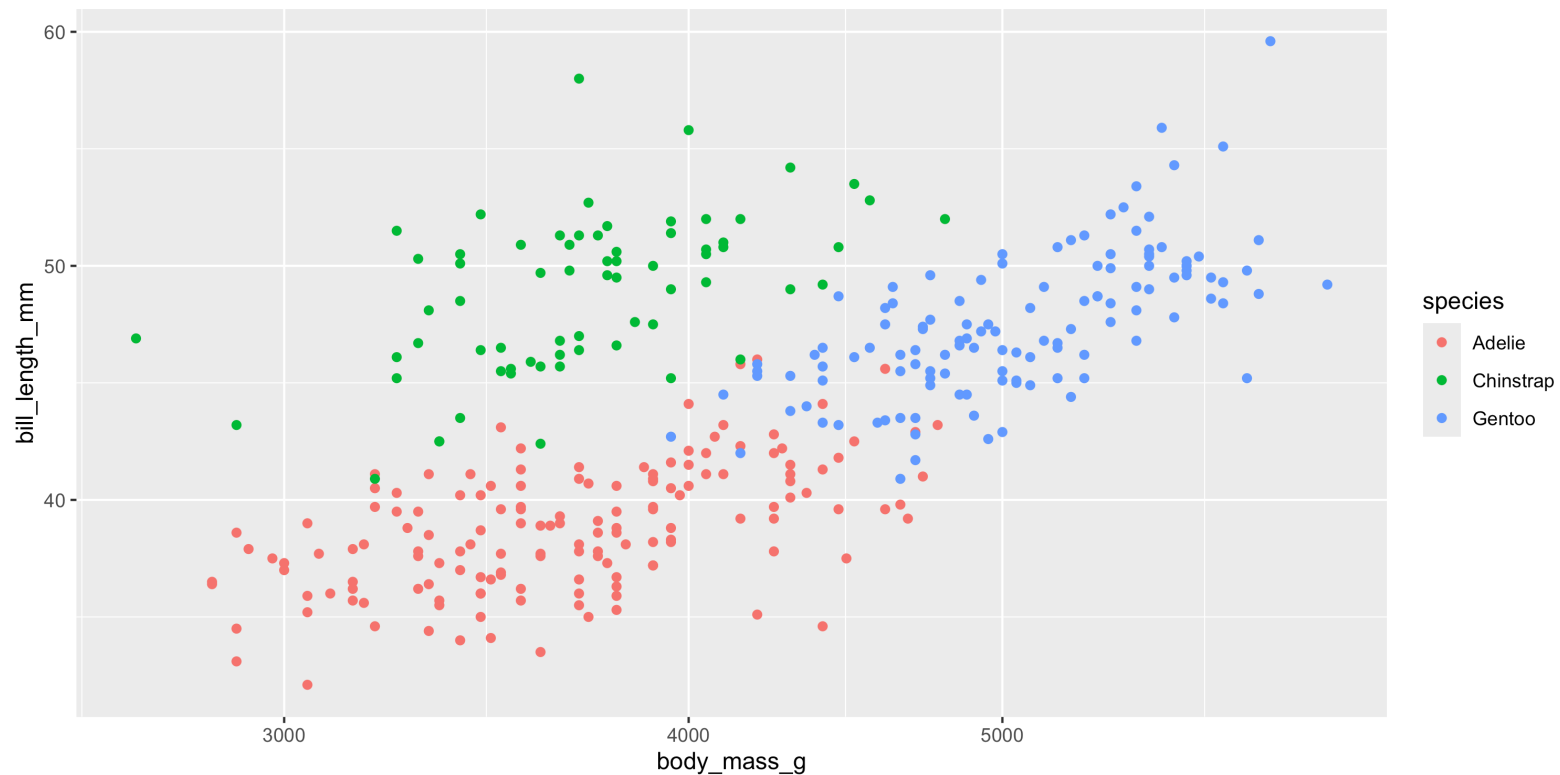
```
1 ggplot(  
2   penguins,  
3   aes(  
4     x = body_mass_g,  
5     y = bill_length_mm  
6   )  
7 ) +  
8 geom_point(  
9   aes(color = species), na.rm = TRUE  
10 ) +  
11 scale_x_continuous(  
12   labels = scales::label_comma(),  
13   breaks = seq(3000, 6000, by = 1000)  
14 ) +  
15 scale_color_viridis_d()
```



The `scales` package provides a variety of useful helper functions for formatting axis labels, e.g. `label_comma()`,

# Log scales

```
1 ggplot(  
2   penguins, aes(x = body_mass_g, y = bill_length_mm)  
3 ) +  
4   geom_point(aes(color = species), na.rm = TRUE) +  
5   scale_x_log10()
```



This can also be achieved via `scale_x_continuous(transform = "log10")` which supports many additional

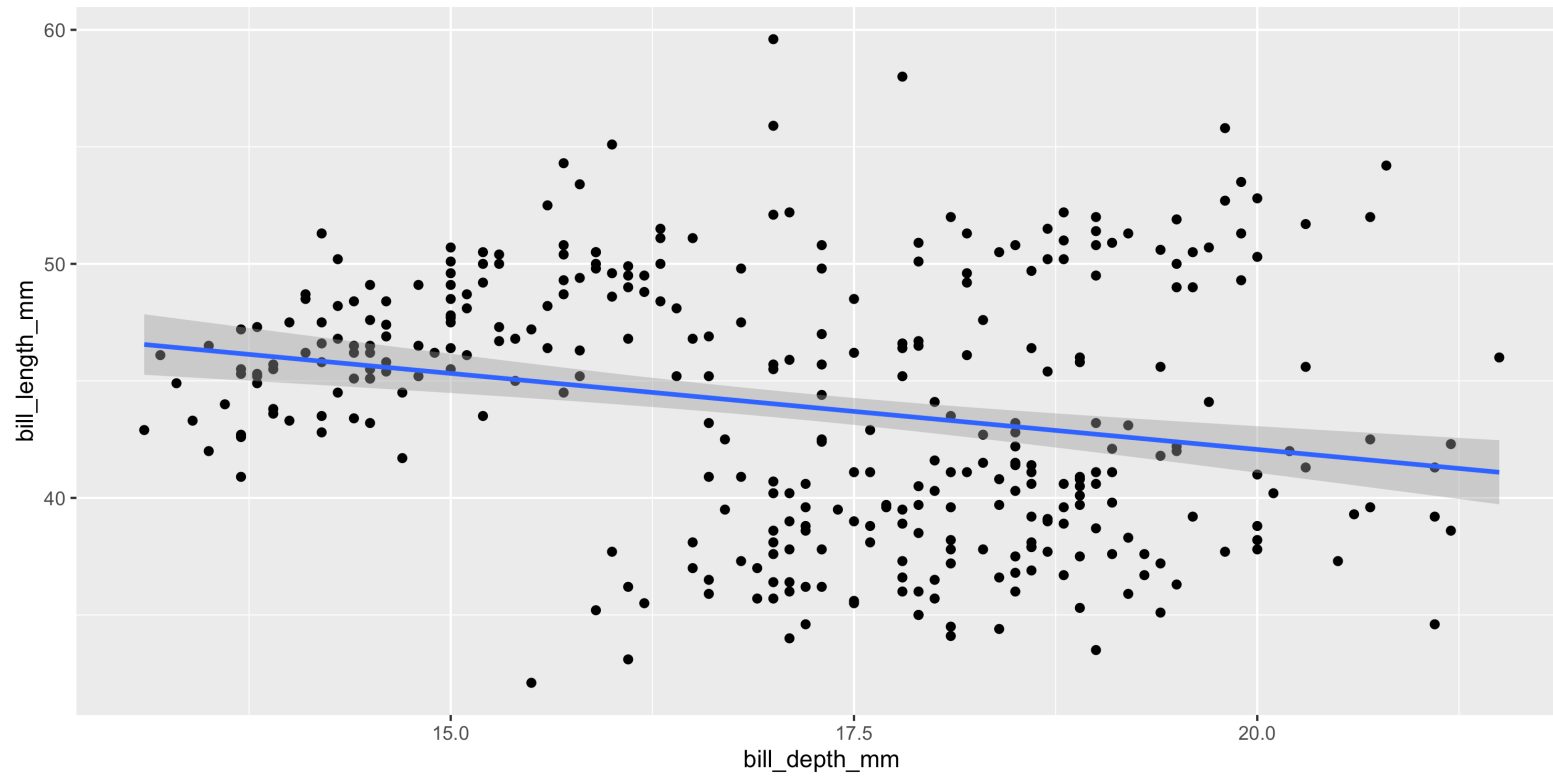
# Statistical Transformations

# Statistical Transformations

Many geoms apply a statistical transformation to the data before plotting — e.g. binning, counting, or fitting a model.

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(na.rm = TRUE) +  
5   geom_smooth(method = "lm", na.rm = TRUE)
```

`geom_smooth()` using formula = 'y ~ x'

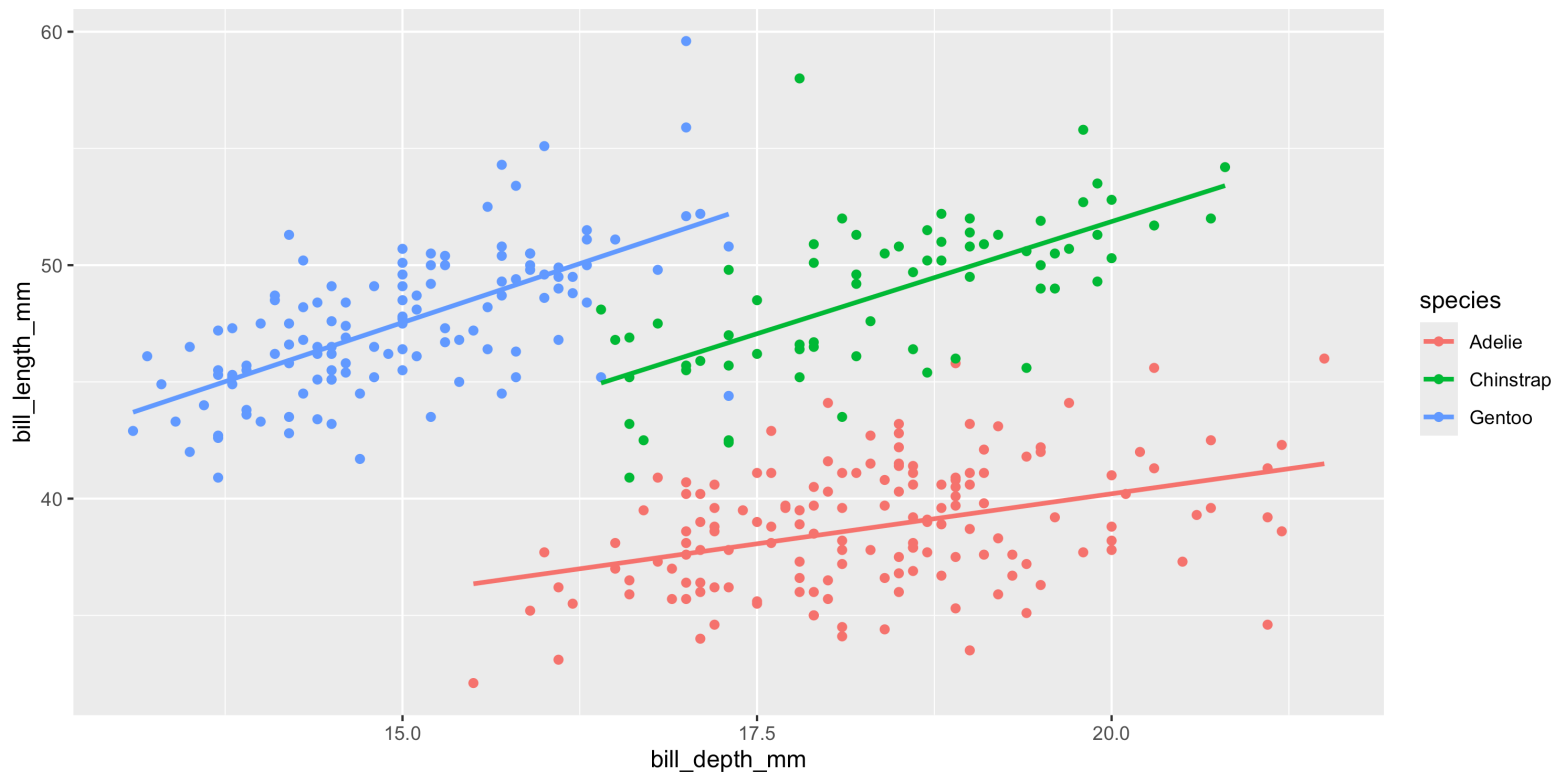


# Stat transformations + aesthetics

Statistical transformations respect aesthetic groupings.

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm, color = species)  
3 ) +  
4   geom_point(na.rm = TRUE) +  
5   geom_smooth(method = "lm", na.rm = TRUE, se = FALSE, )
```

`geom_smooth()` using formula = 'y ~ x'



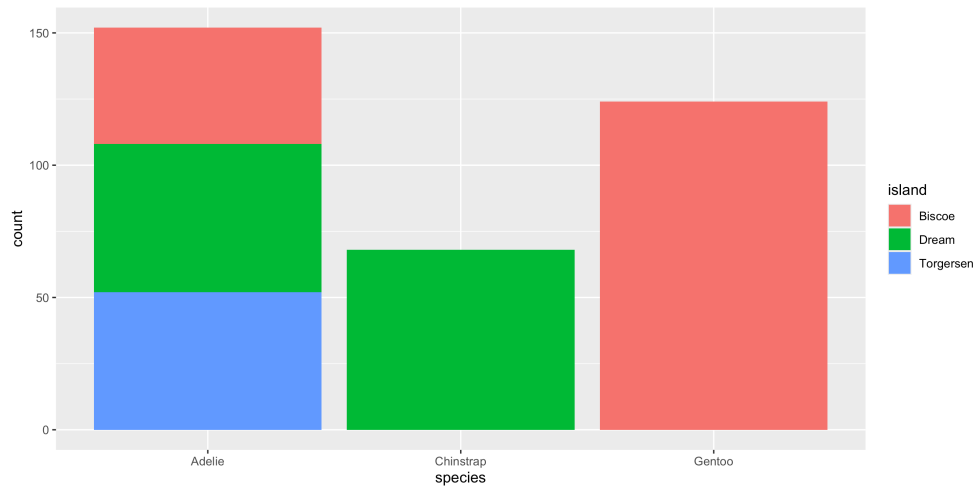
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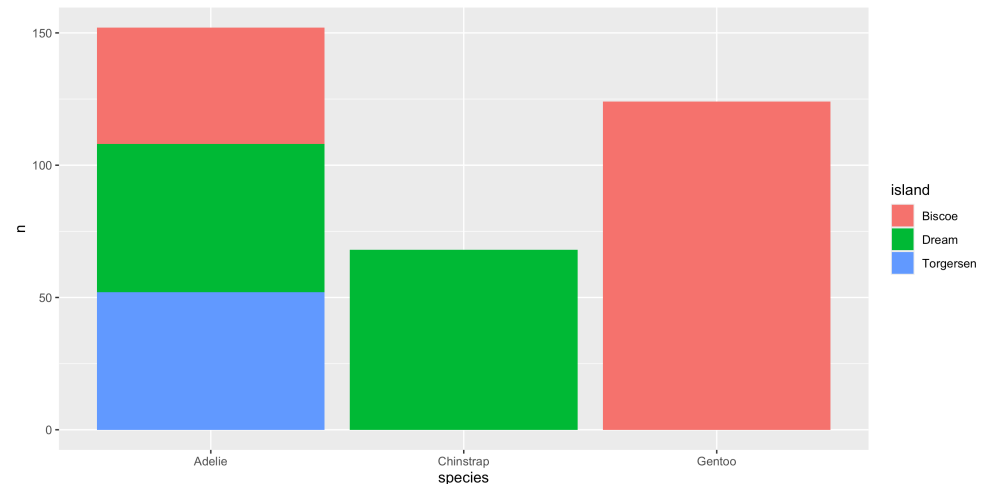
# Implicit stat transformations

Some geoms imply a stat — `geom_bar()` uses `stat = "count"` by default, so it only needs an `x` aesthetic.

```
1 ggplot(  
2   penguins, aes(x = species, fill = island)  
3 ) +  
4   geom_bar()
```



```
1 penguins |>  
2   count(species, island) |>  
3   ggplot(  
4     aes(x = species, y = n, fill = island)  
5   ) +  
6     geom_bar(stat = "identity")
```



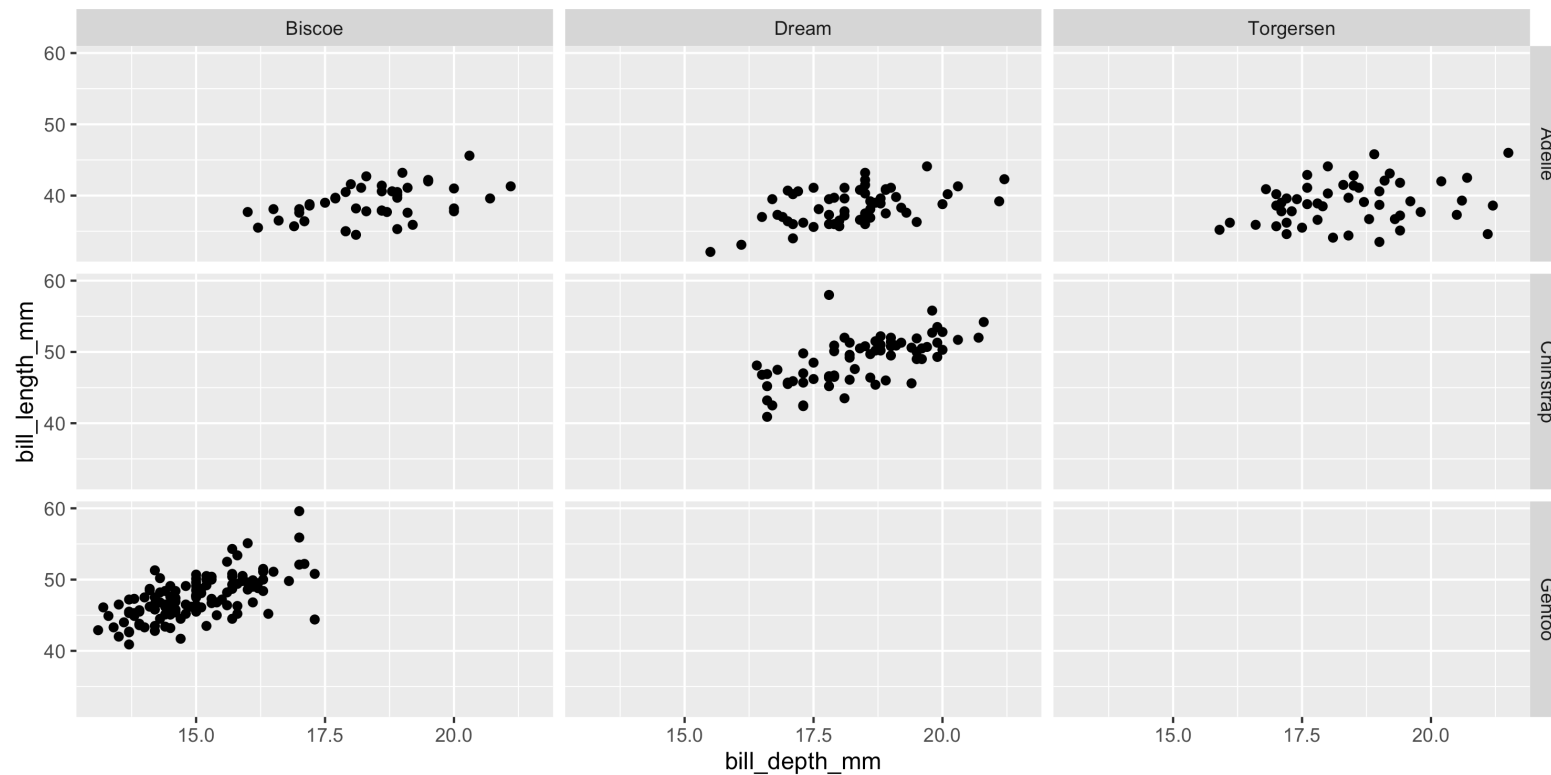
# Faceting

# Faceting

- Smaller plots that display different subsets of the data
- Useful for exploring conditional relationships and large data
- Sometimes referred to as “small multiples”

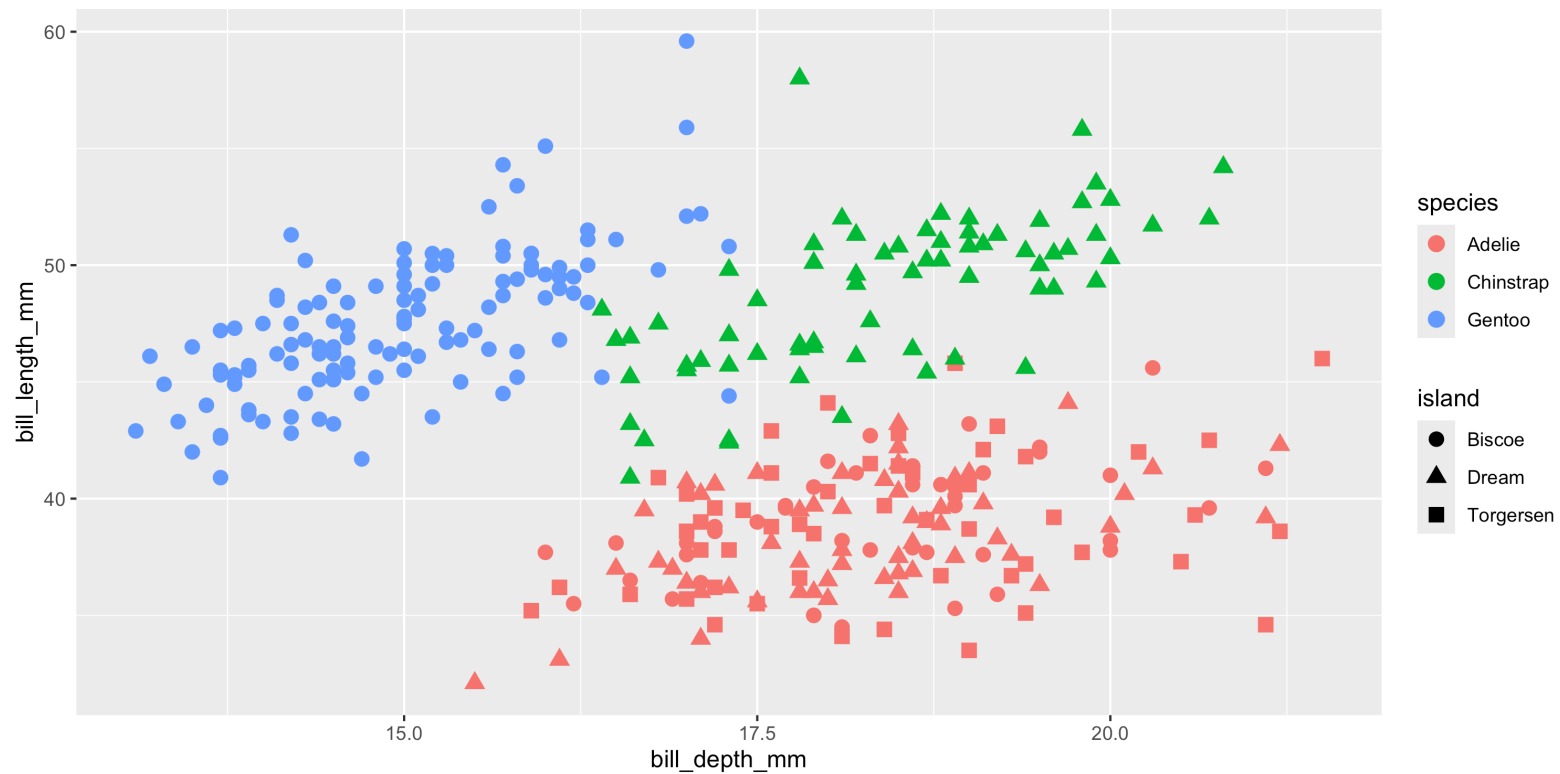
# facet\_grid

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(na.rm = TRUE) +  
5   facet_grid(  
6     species ~ island  
7   )
```



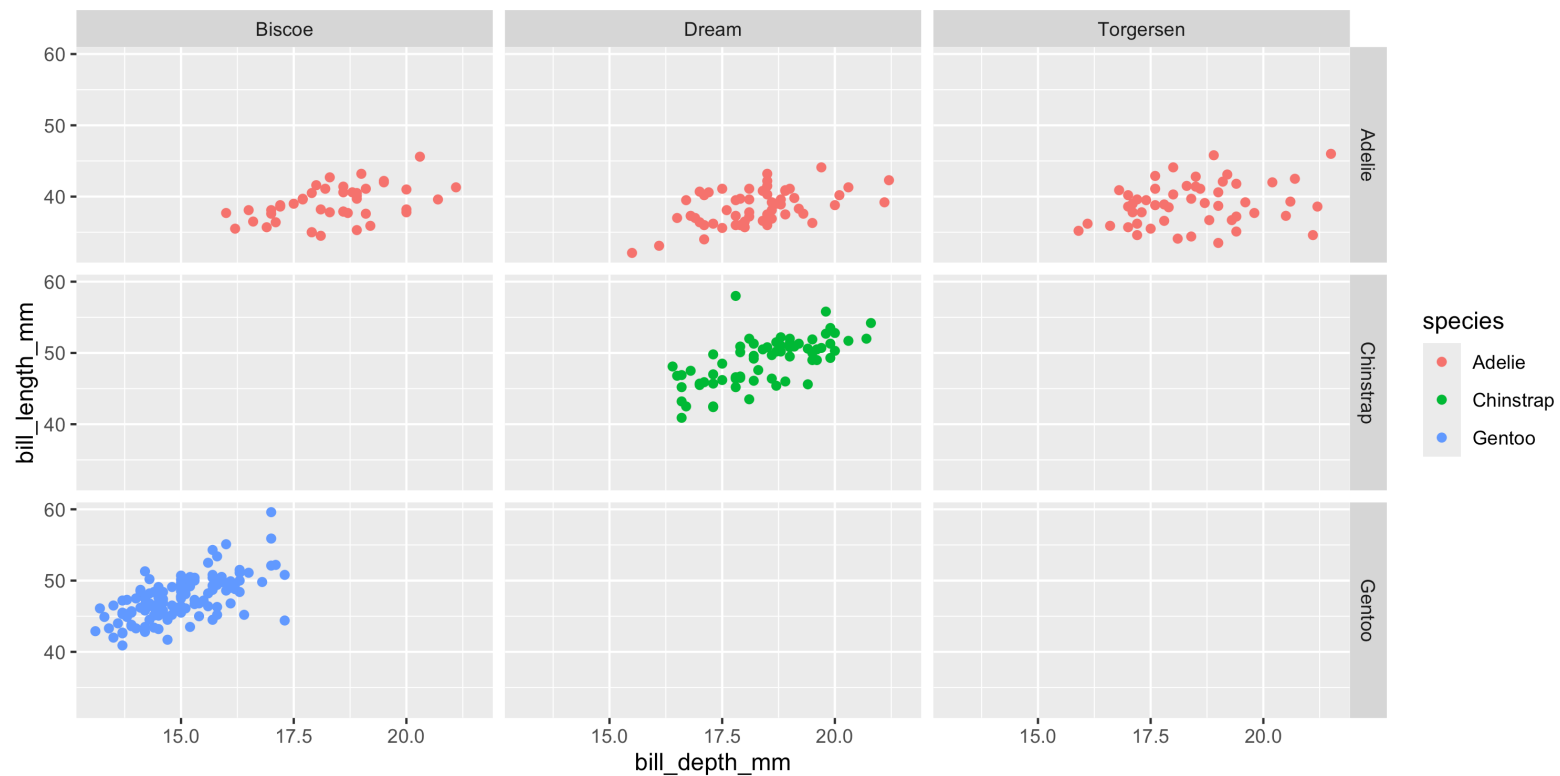
# Compare with ...

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(  
5     aes(color = species, shape = island), na.rm = TRUE, size = 3  
6   )
```



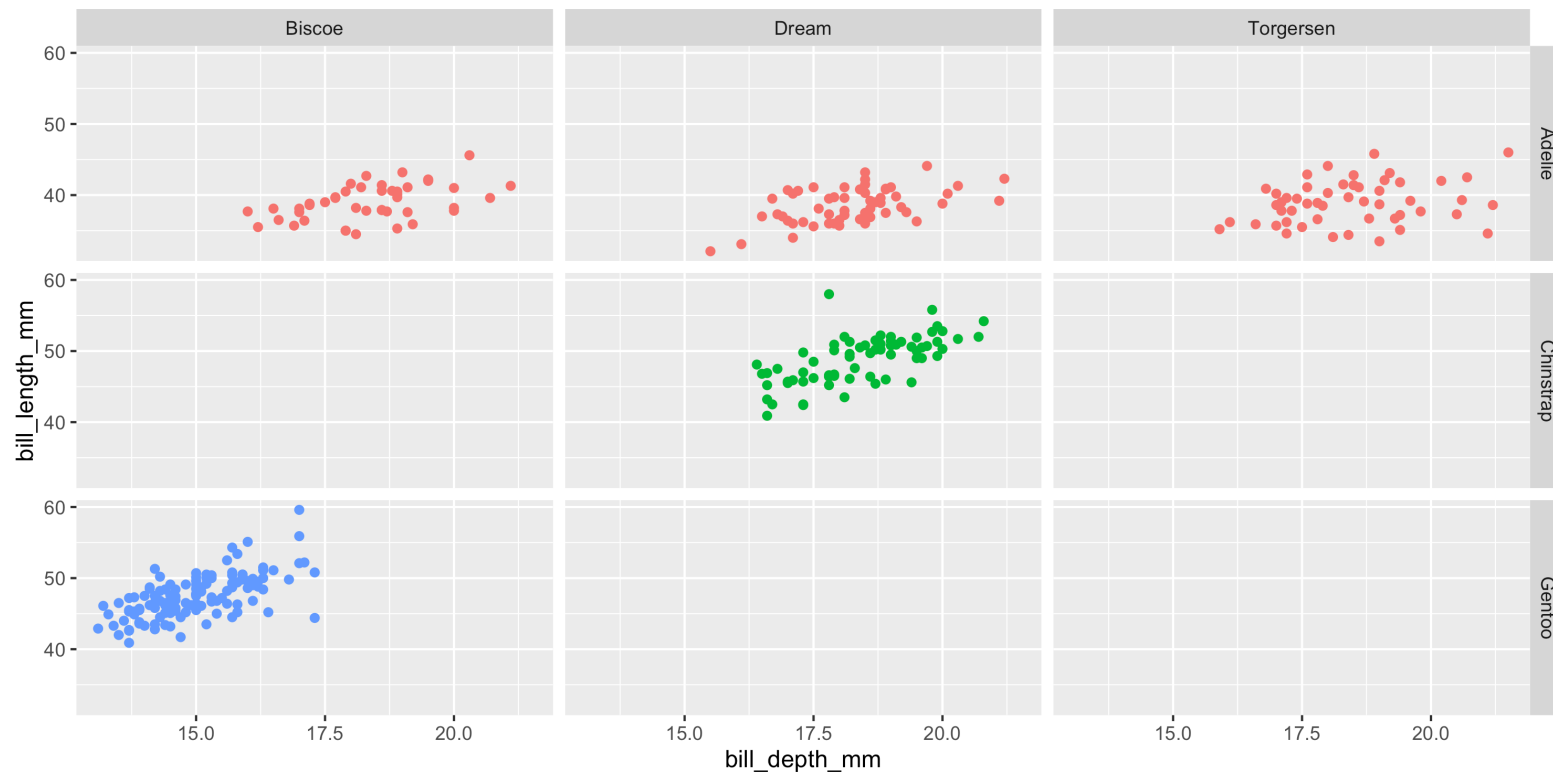
# Faceting and color

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm, color = species)  
3 ) +  
4   geom_point(na.rm = TRUE) +  
5   facet_grid(species ~ island)
```



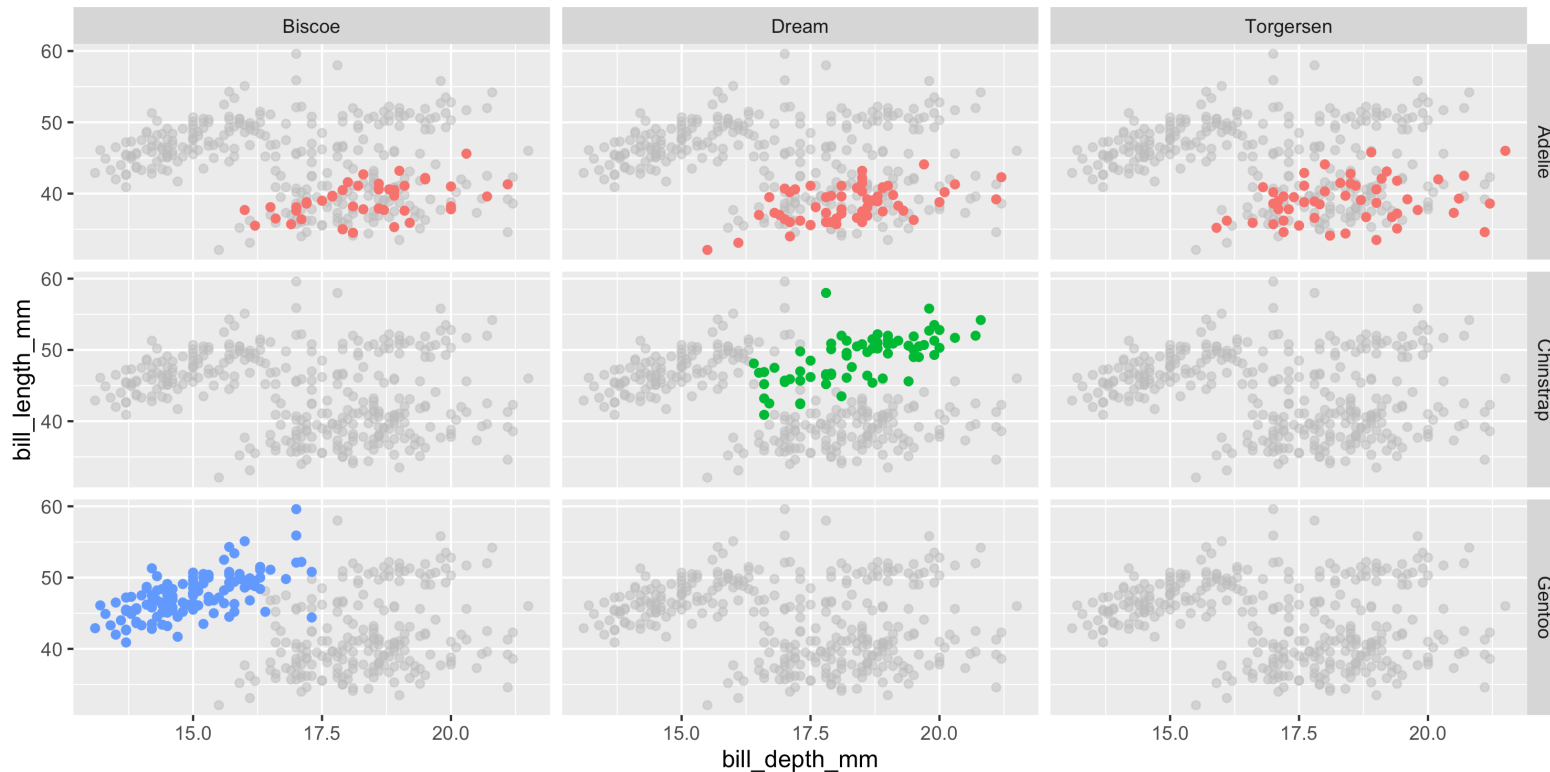
# Hiding legend elements

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm, color = species)  
3 ) +  
4   geom_point(na.rm = TRUE) +  
5   facet_grid(species ~ island) +  
6   guides(color = "none")
```



# Facet layout - context

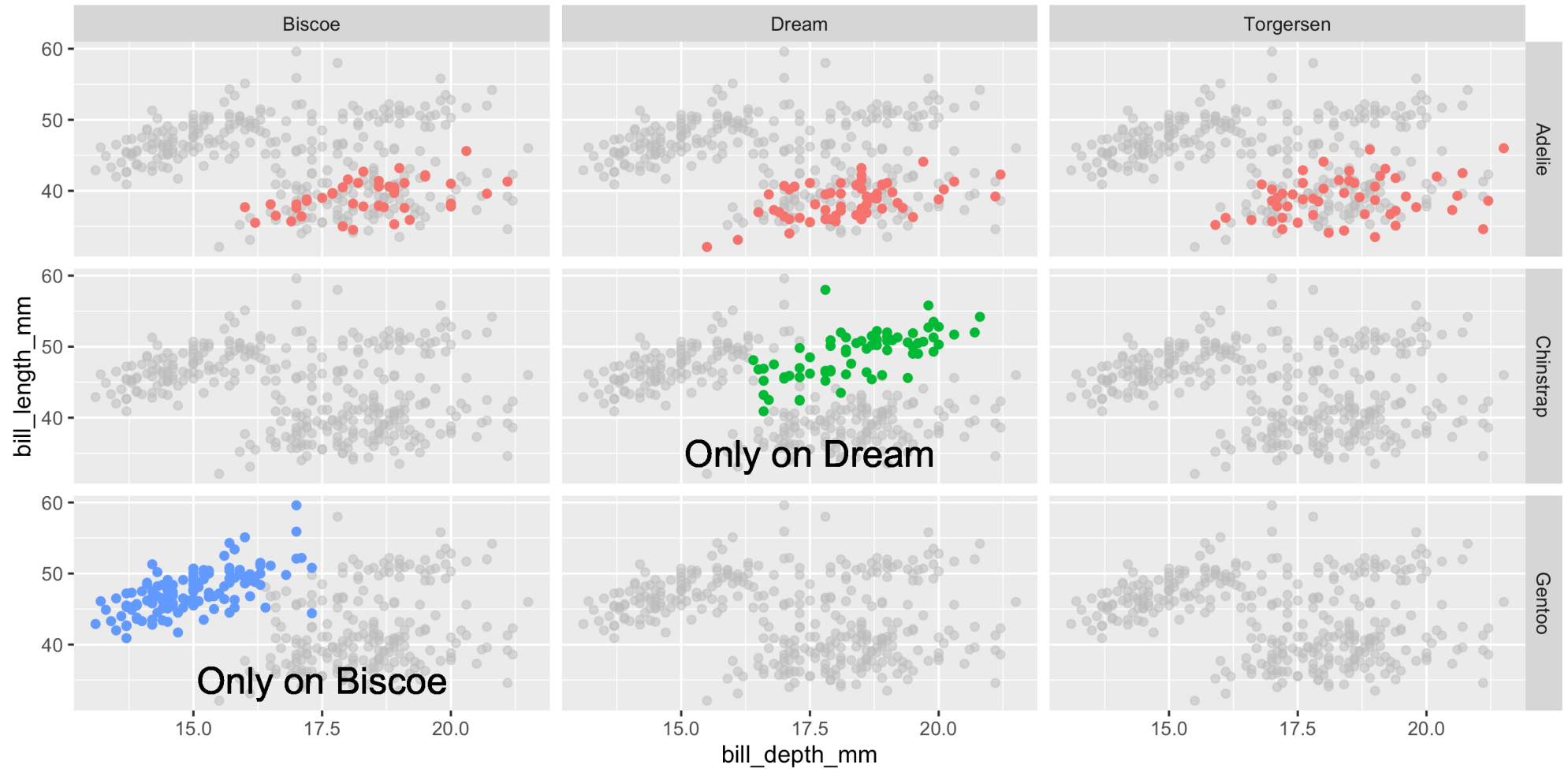
```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm, color = species)  
3 ) +  
4   geom_point(color = "grey", alpha = 0.5, na.rm = TRUE, layout = "fixed") +  
5   geom_point(na.rm = TRUE) +  
6   facet_grid(species ~ island) +  
7   guides(color = "none")
```



# Facet layout - annotation

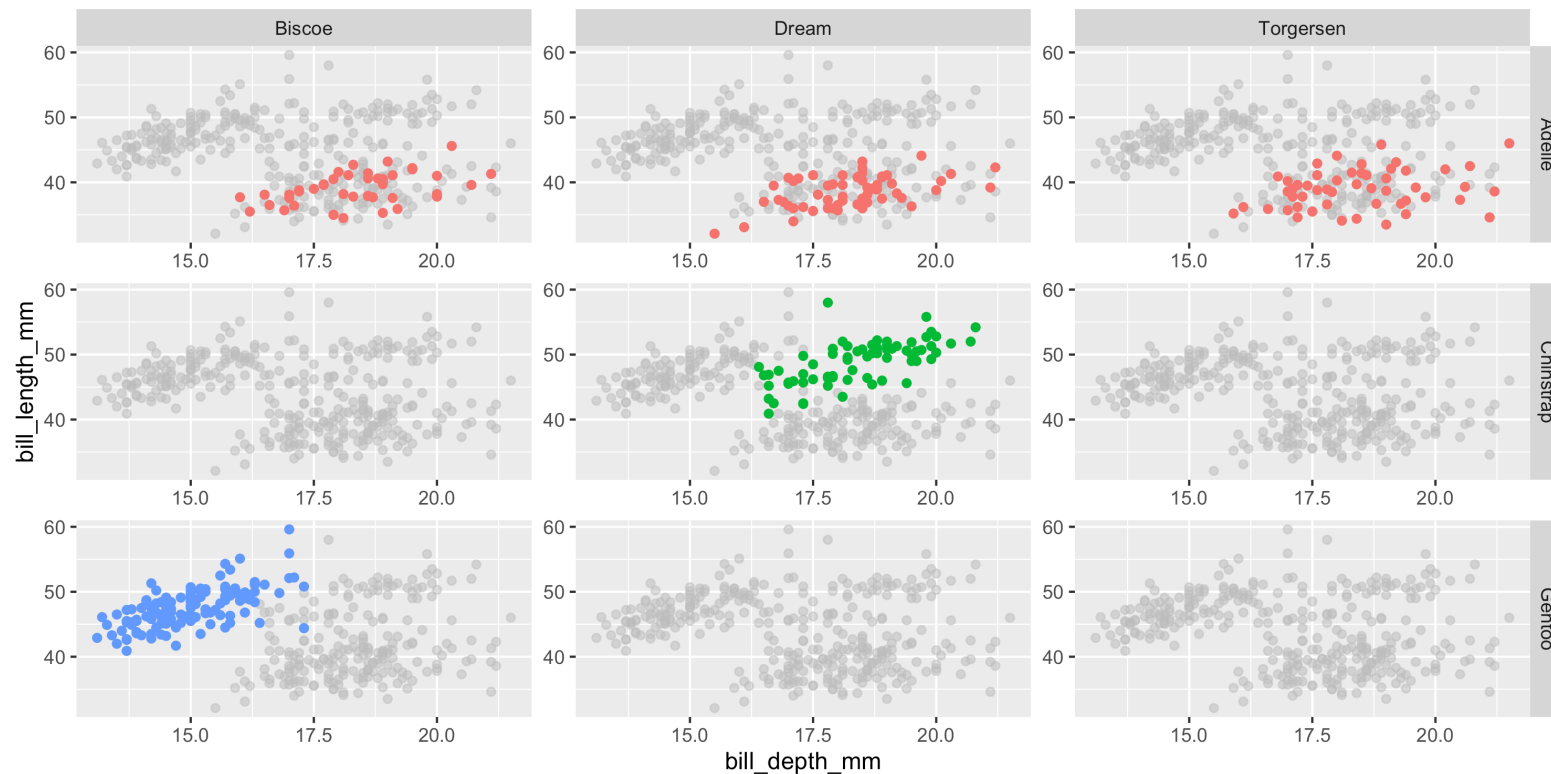
```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm, color = species)  
3 ) +  
4   geom_point(color = "grey", alpha = 0.5, na.rm = TRUE, layout = "fixed") +  
5   geom_point(na.rm = TRUE) +  
6   facet_grid(species ~ island) +  
7   guides(color = "none") +  
8   geom_text(  
9     x = 17.5, y = 35, label = "Only on Dream", size = 6, color = "black",  
10    layout = 5  
11 ) +  
12   geom_text(  
13     x = 17.5, y = 35, label = "Only on Biscoe", size = 6, color = "black",  
14     layout = 7  
15 )
```

# Facet layout - annotation



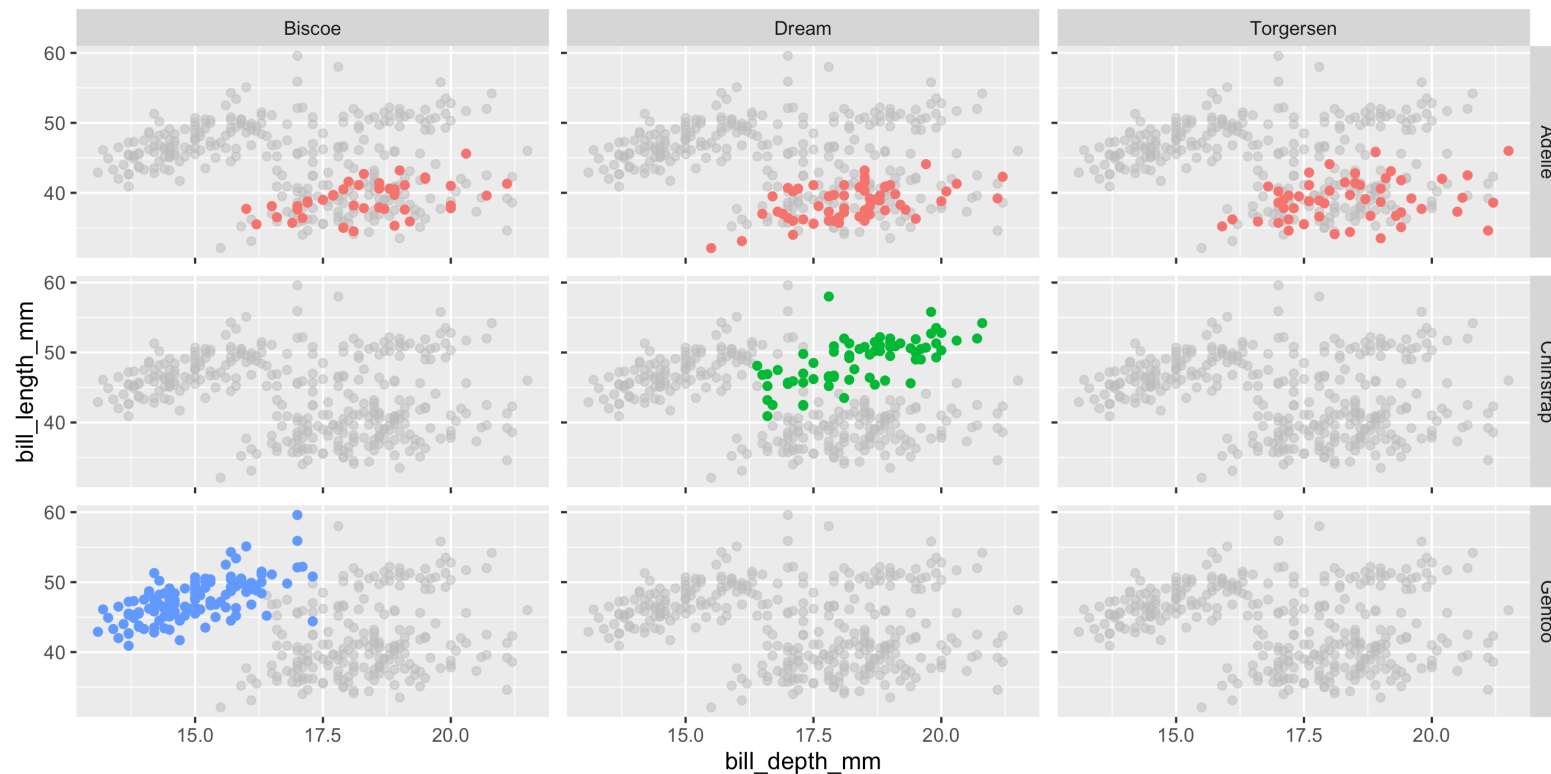
# Facet axes

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm, color = species)  
3 ) +  
4   geom_point(color = "grey", alpha = 0.5, na.rm = TRUE, layout = "fixed") +  
5   geom_point(na.rm = TRUE) +  
6   facet_grid(species ~ island, axes = "all") +  
7   guides(color = "none")
```



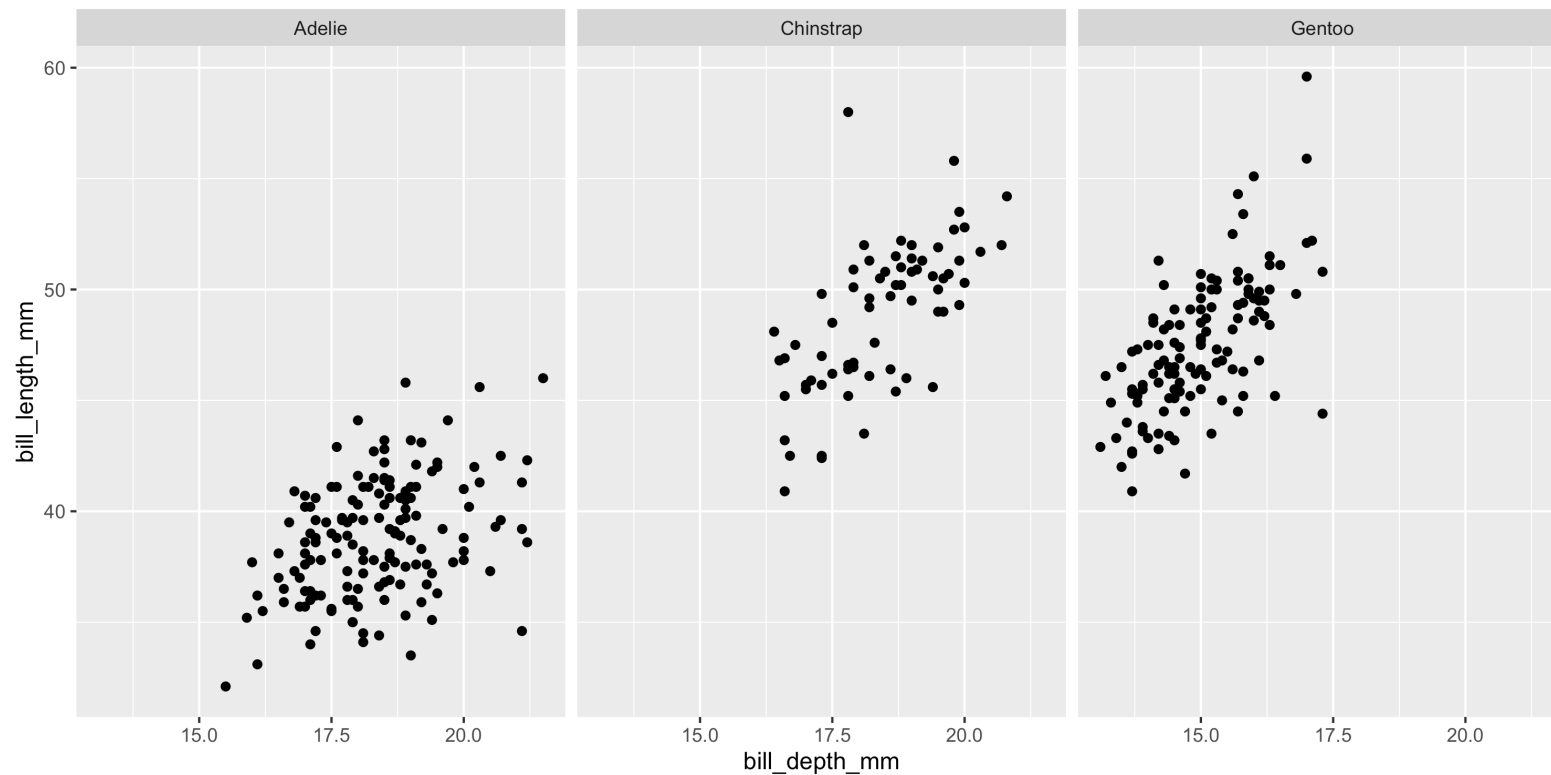
# Facet axes - labels

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm, color = species)  
3 ) +  
4   geom_point(color = "grey", alpha = 0.5, na.rm = TRUE, layout = "fixed") +  
5   geom_point(na.rm = TRUE) +  
6   facet_grid(species ~ island, axes = "all", axis.labels = "margins") +  
7   guides(color = "none")
```



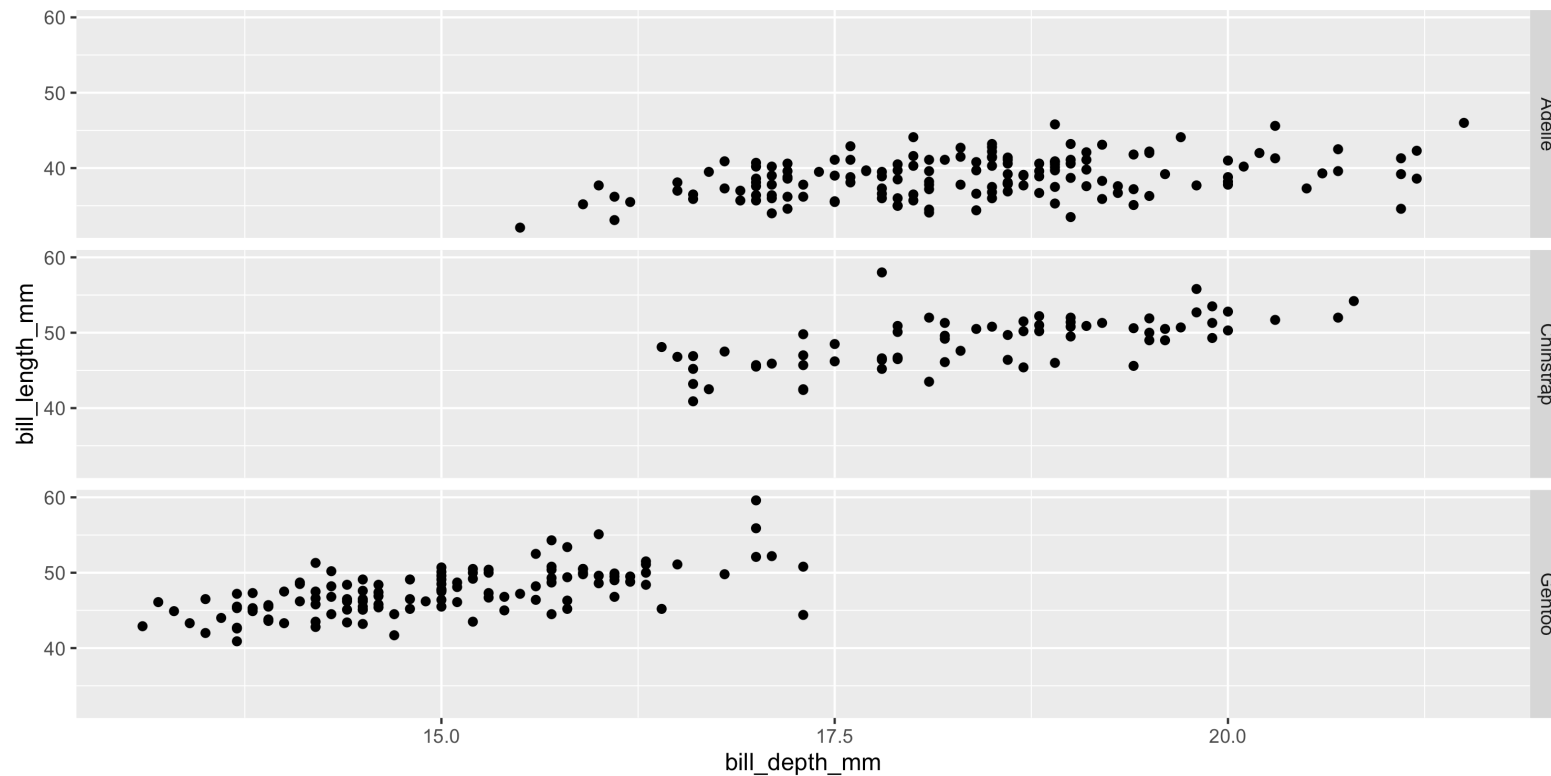
# facet\_grid (columns)

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(na.rm = TRUE) +  
5   facet_grid(~ species)
```



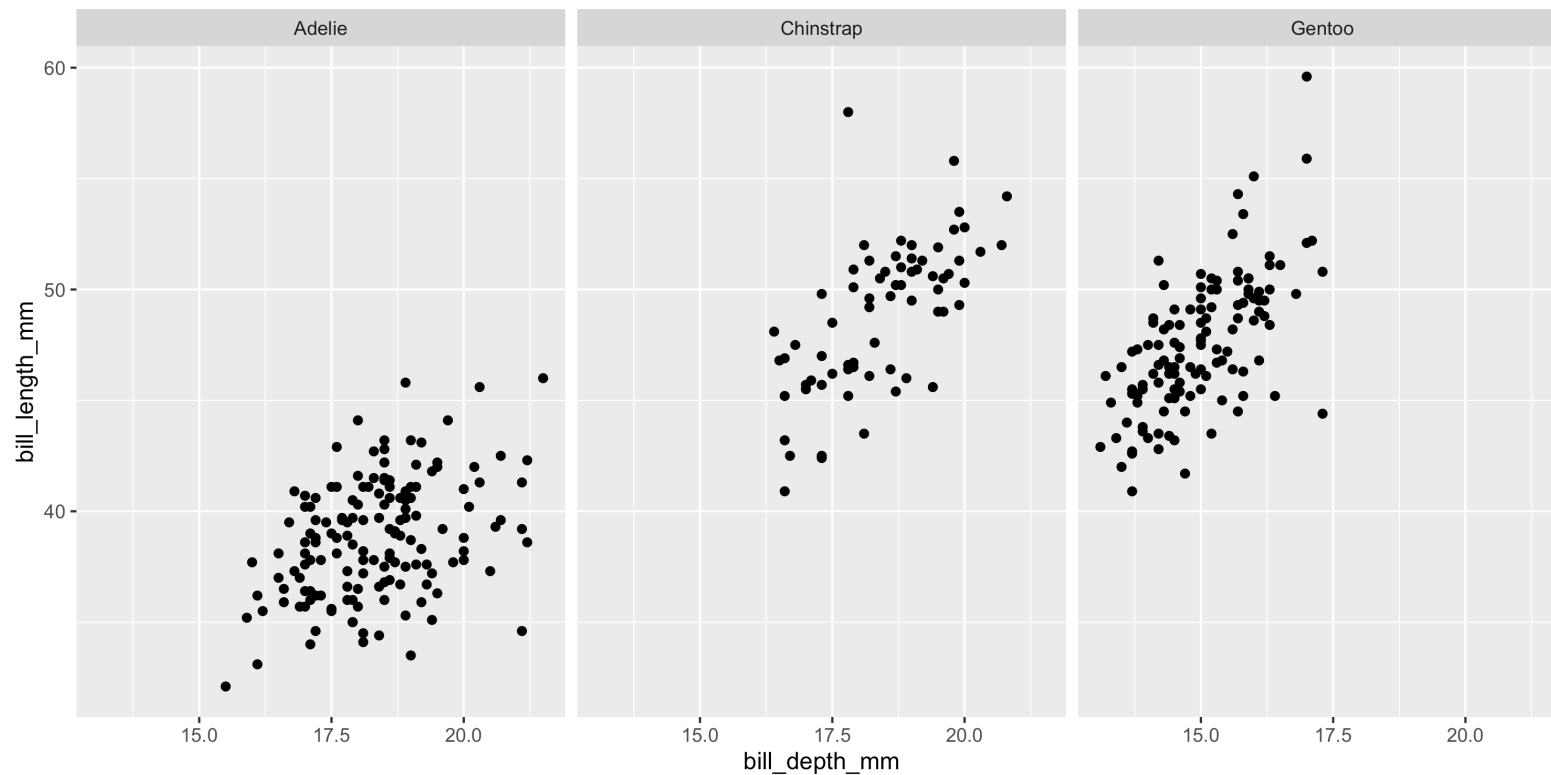
# facet\_grid (rows)

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(na.rm = TRUE) +  
5   facet_grid(species ~ .)
```



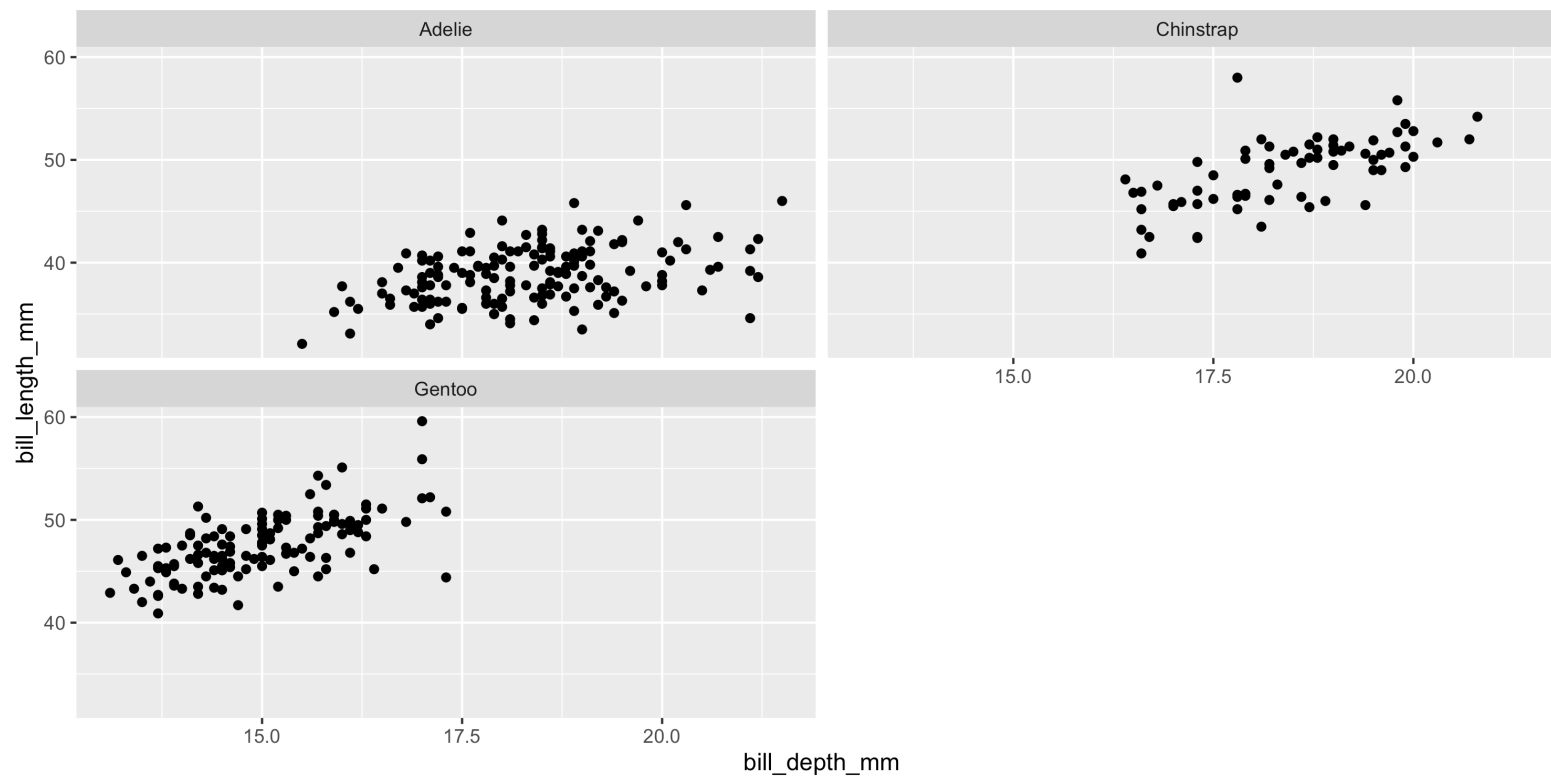
# facet\_wrap

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(na.rm = TRUE) +  
5   facet_wrap(~ species)
```



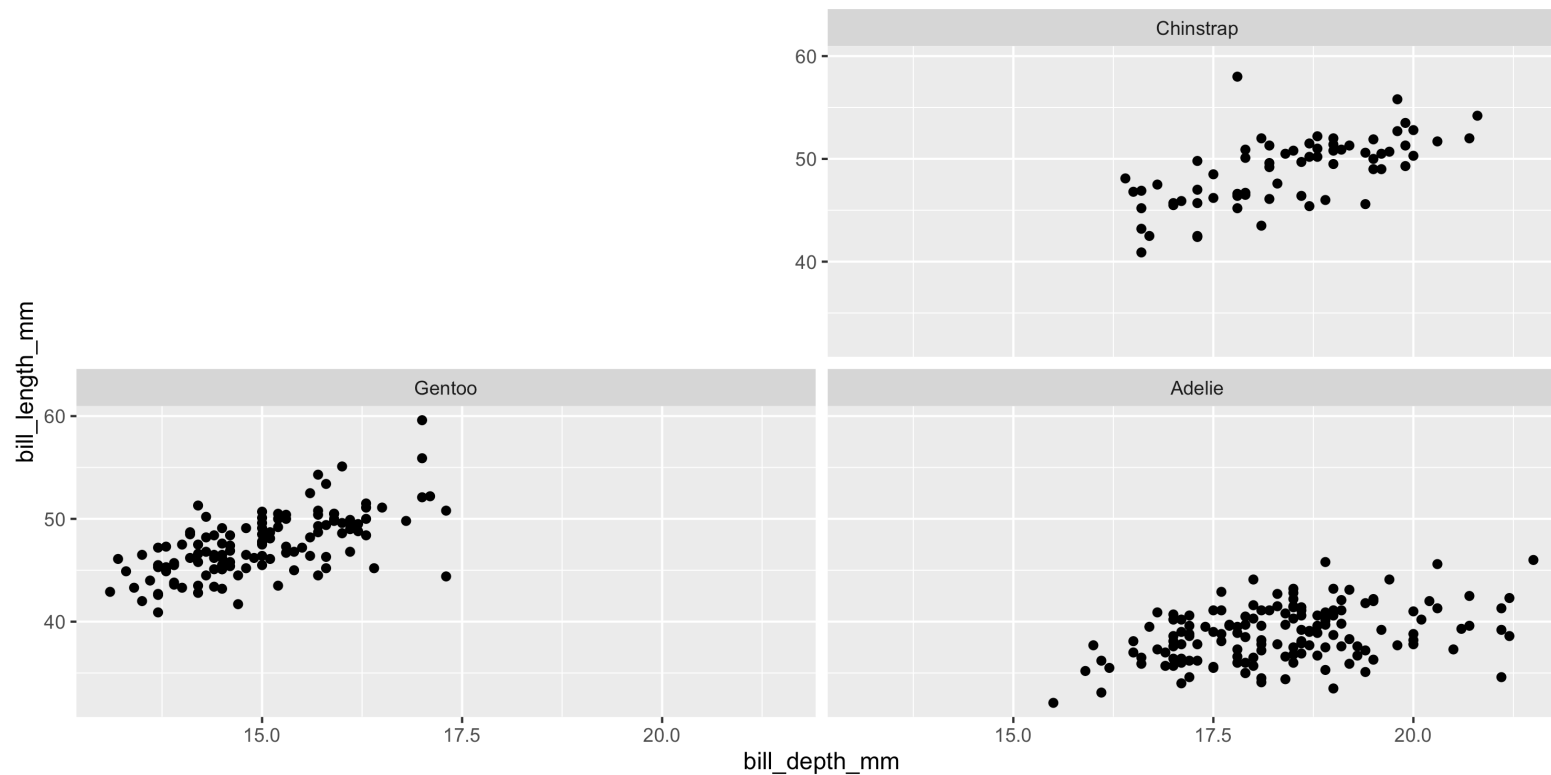
# facet\_wrap

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(na.rm = TRUE) +  
5   facet_wrap(~ species, ncol = 2)
```



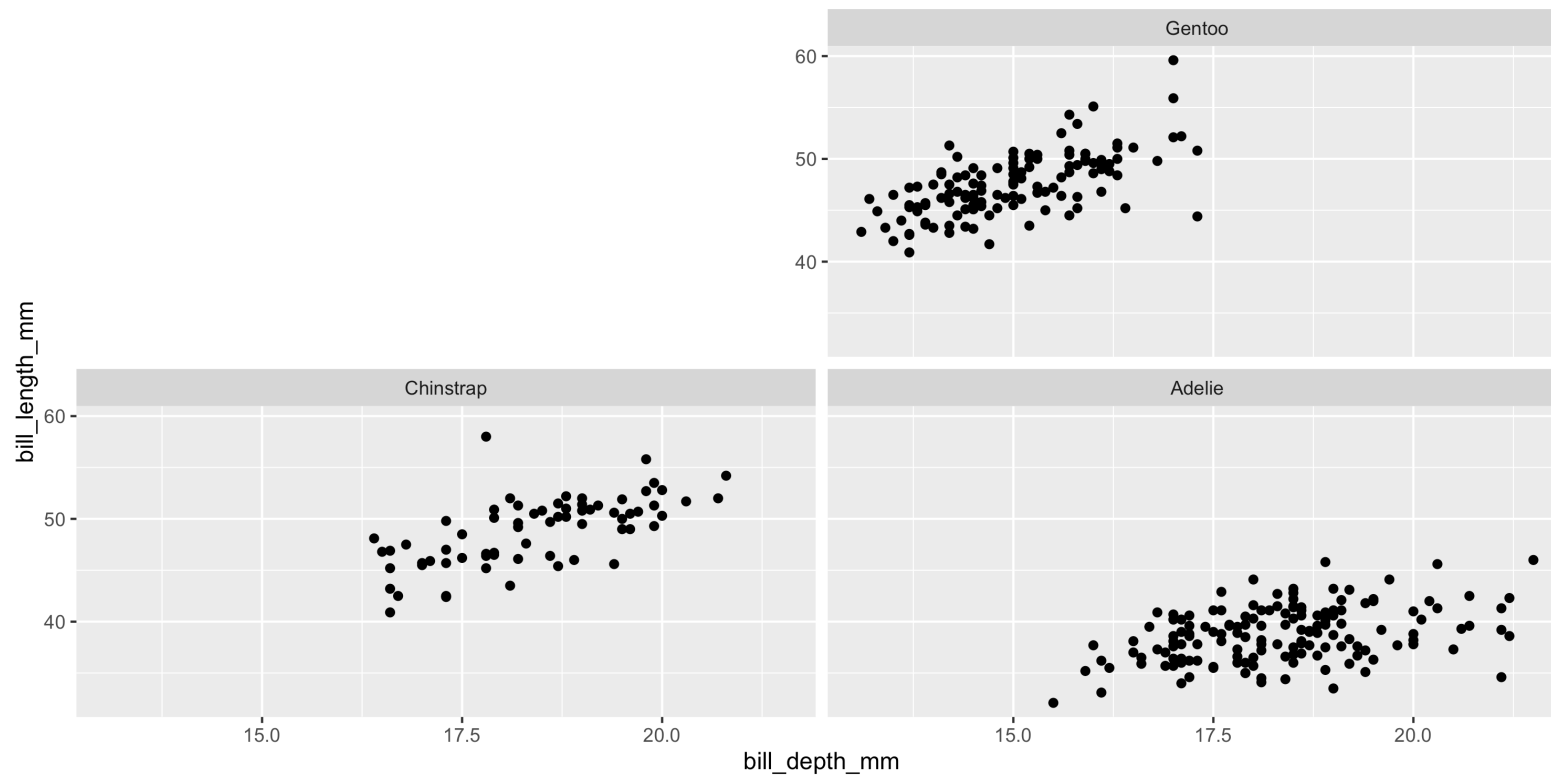
# facet\_wrap - direction

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(na.rm = TRUE) +  
5   facet_wrap(~ species, ncol = 2, dir = "br")
```



# facet\_wrap - direction

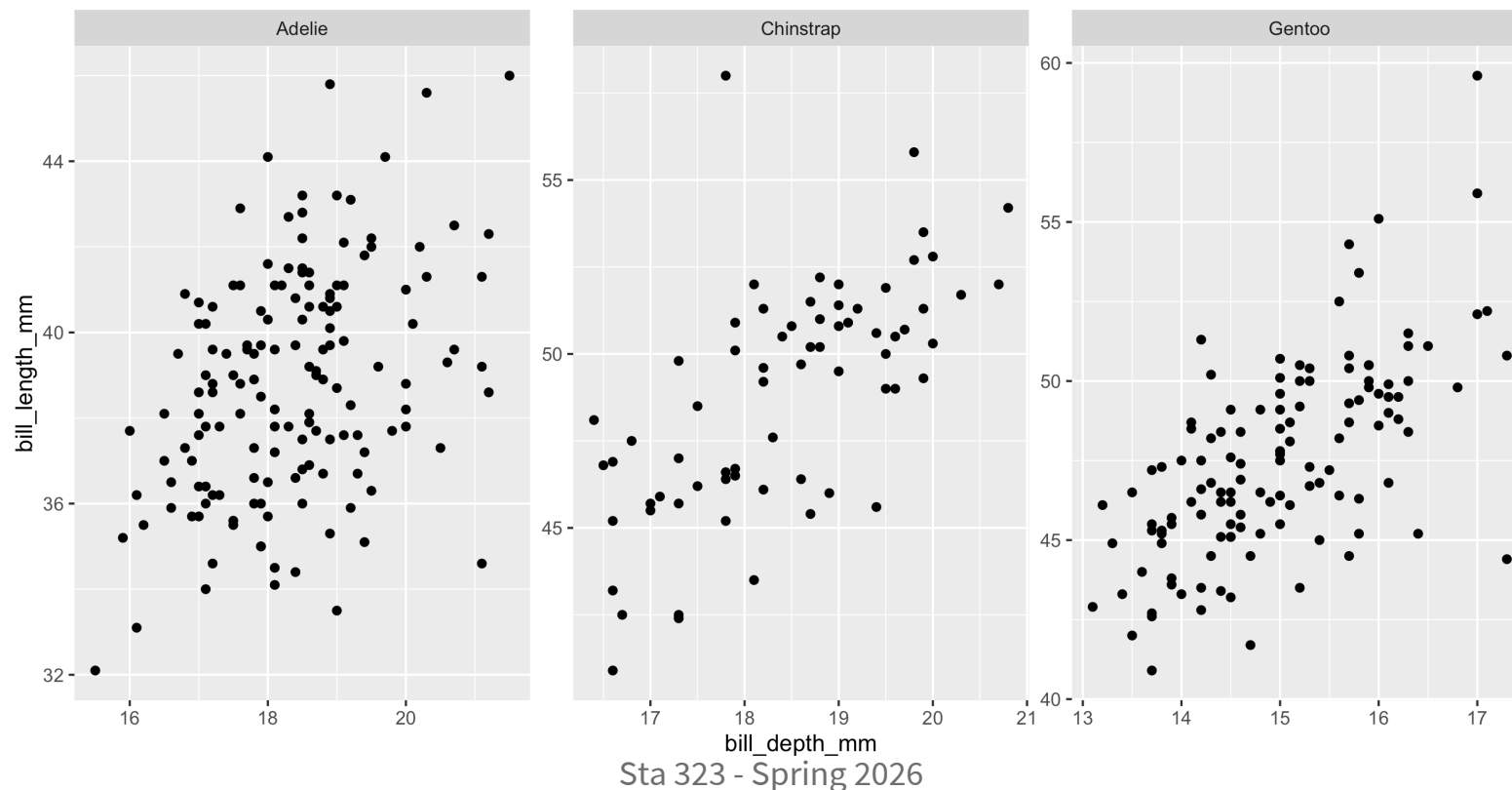
```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(na.rm = TRUE) +  
5   facet_wrap(~ species, ncol = 2, dir = "rb")
```



# facet\_wrap - free scales

By default, all facets share the same axis limits. Use `scales` to let axes vary across panels.

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm)  
3 ) +  
4   geom_point(na.rm = TRUE) +  
5   facet_wrap(~ species, scales = "free")
```

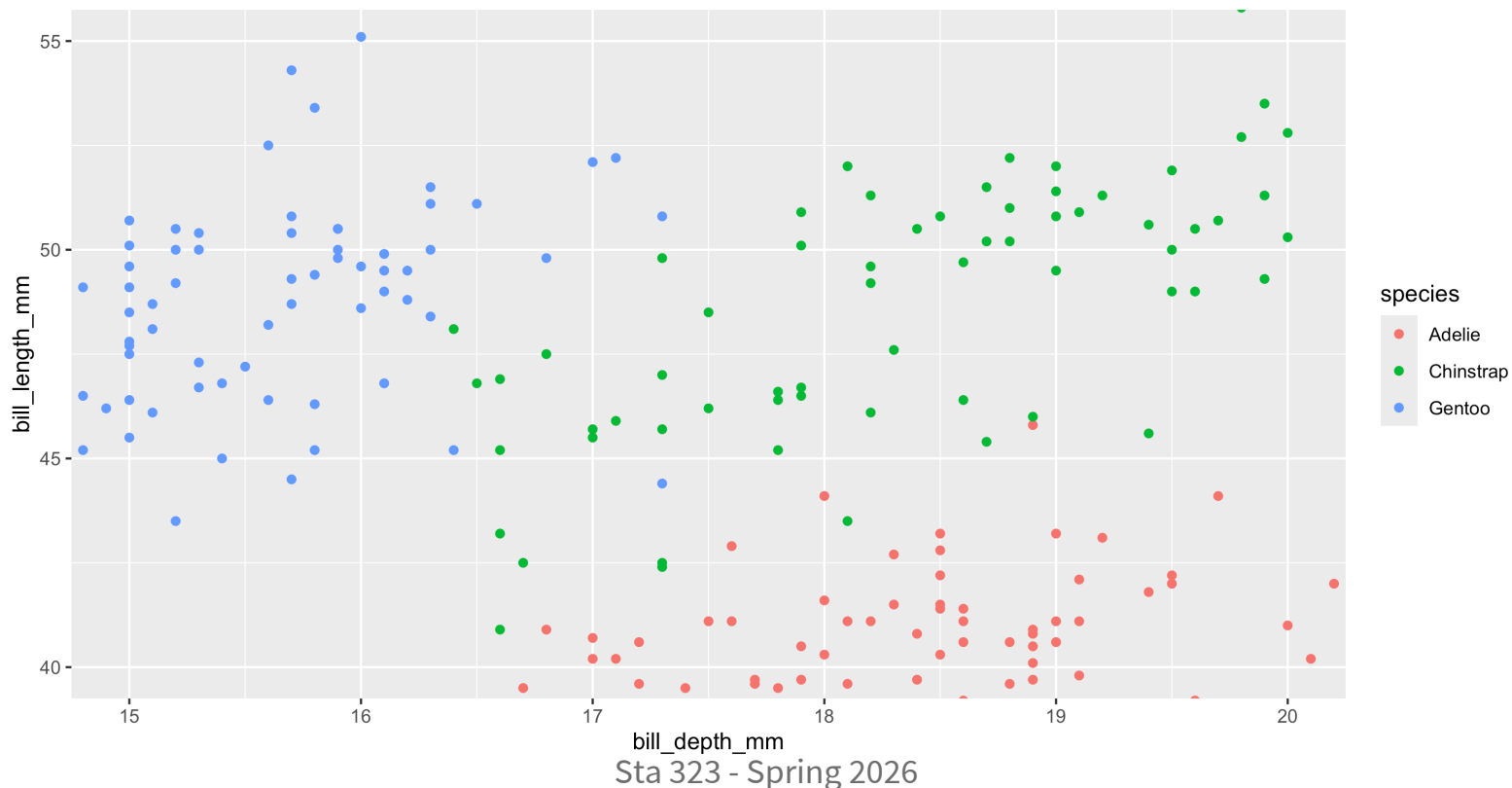


# Coordinate Systems

# coord\_cartesian - zooming

`coord_cartesian()` zooms into the plot without dropping data — unlike setting `scale` limits which removes points before stat computations.

```
1 ggplot(  
2   penguins, aes(x = bill_depth_mm, y = bill_length_mm, color = species)  
3 ) +  
4   geom_point(na.rm = TRUE) +  
5   coord_cartesian(xlim = c(15, 20), ylim = c(40, 55))
```

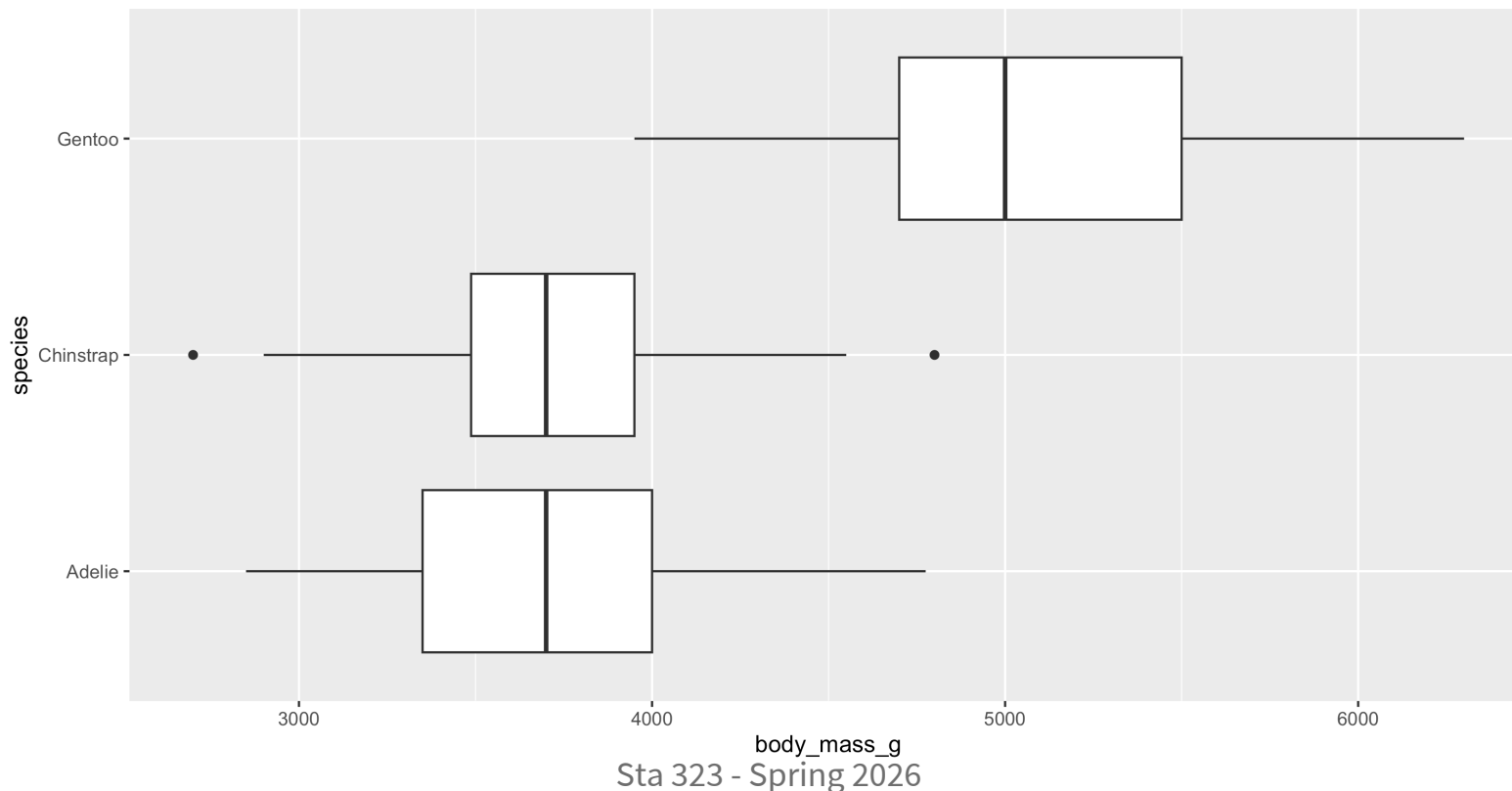




# coord\_flip

`coord_flip()` swaps the x and y axes — useful for making horizontal bar charts or boxplots more readable.

```
1 ggplot(  
2   penguins, aes(x = species, y = body_mass_g)  
3 ) +  
4   geom_boxplot(na.rm = TRUE) +  
5   coord_flip()
```

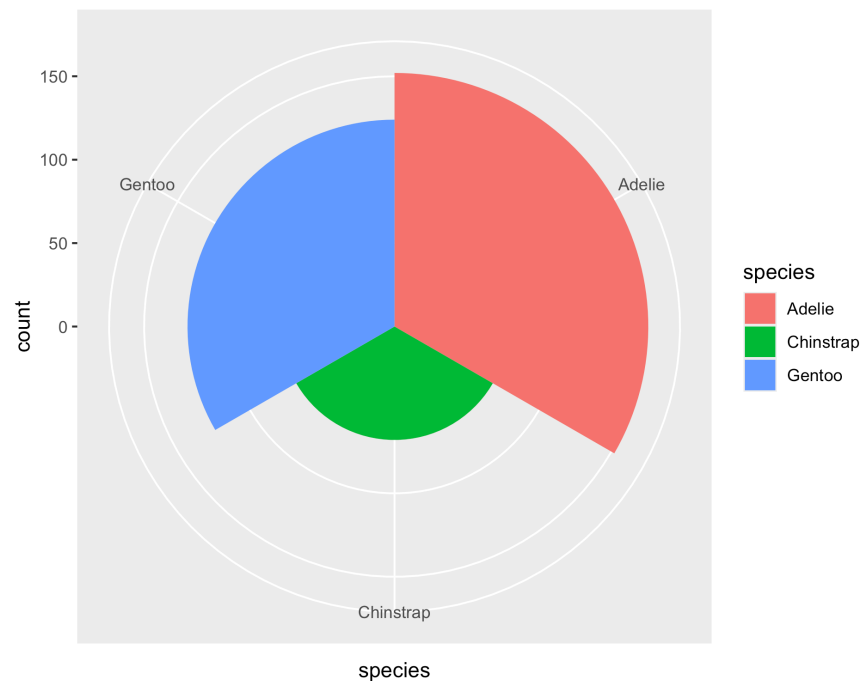




# coord\_polar

`coord_polar()` maps position onto a circular coordinate system.

```
1 ggplot(  
2   penguins |> dplyr::filter(!is.na(species)),  
3   aes(x = species, fill = species)  
4 ) +  
5   geom_bar(width = 1) +  
6   coord_polar()
```



# Learning more

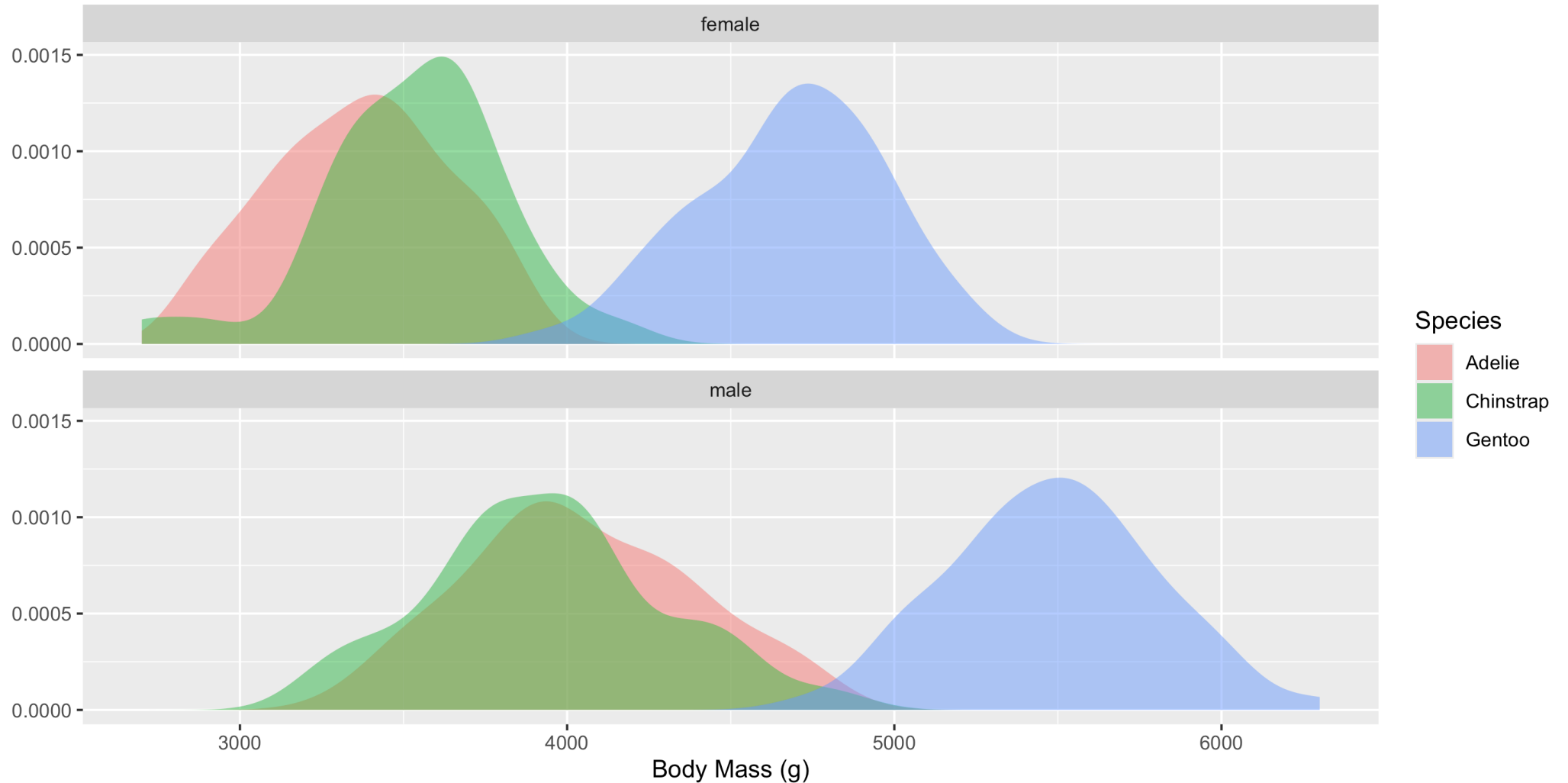
# geom tour

[ggplot2.tidyverse.org/reference/index.html](https://ggplot2.tidyverse.org/reference/index.html)

# Exercises

# Exercise 1

Recreate, as faithfully as possible, the following plot using ggplot2 and the `penguins` data.



# Exercise 2

Recreate, as faithfully as possible, the following plot from the `palmerpenguins` package README in `ggplot2`.

