

Making Data Speak: Principles and Practice of Scientific Visualization

Part 1: Code-Along

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Some useful keyboard shortcuts:

- New code chunk: Ctrl+Alt+i // Cmd+Option+i
- Comment or uncomment: Ctrl+Shift+C // Cmd+Shift+C
- Run current selected code line: Ctrl+Enter // Cmd+Return
- Select Word: Ctrl+Shift+Left/Right // Option+Shift+Left/Right
- Select to Line Start: Alt+Shift+Left // Cmd+Shift+Left
- Select to Line End: Alt+Shift+Right // Cmd+Shift+Right

Attach Packages

```
# Core
library(ggplot2)
library(here)

# Data import + wrangling
library(readr)
library(dplyr)
library(tidyr)
library(stringr)
library(lubridate)
library(forcats)
library(janitor)

# Plot helpers
library(scales)
library(patchwork)
```

0. Read in Data

```
crossings <- readr::read_csv(here("SDSU Data Viz 2026/crossings.csv")) %>%
  janitor::clean_names()

penguins <- readr::read_csv(here("SDSU Data Viz 2026/penguins.csv")) %>%
  janitor::clean_names()
```

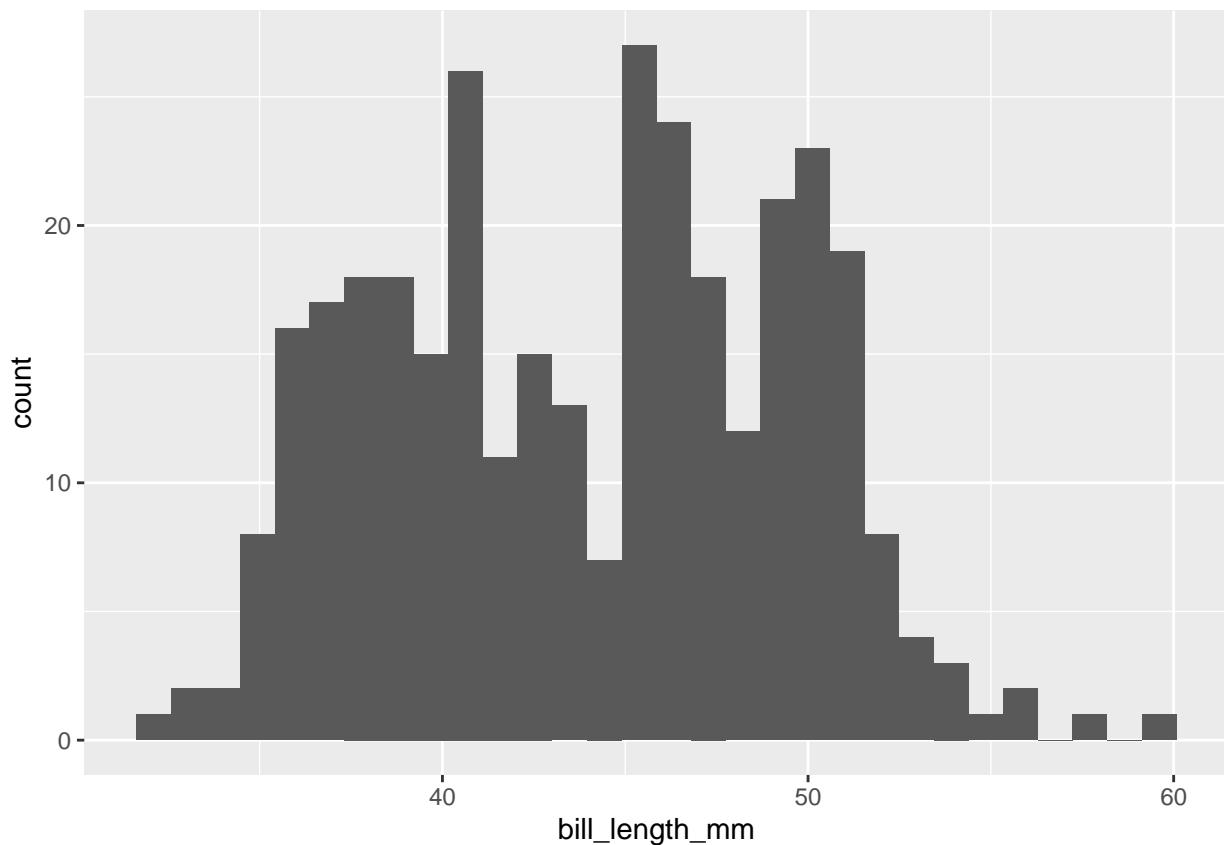
1. Look at data

1.1 summary stats

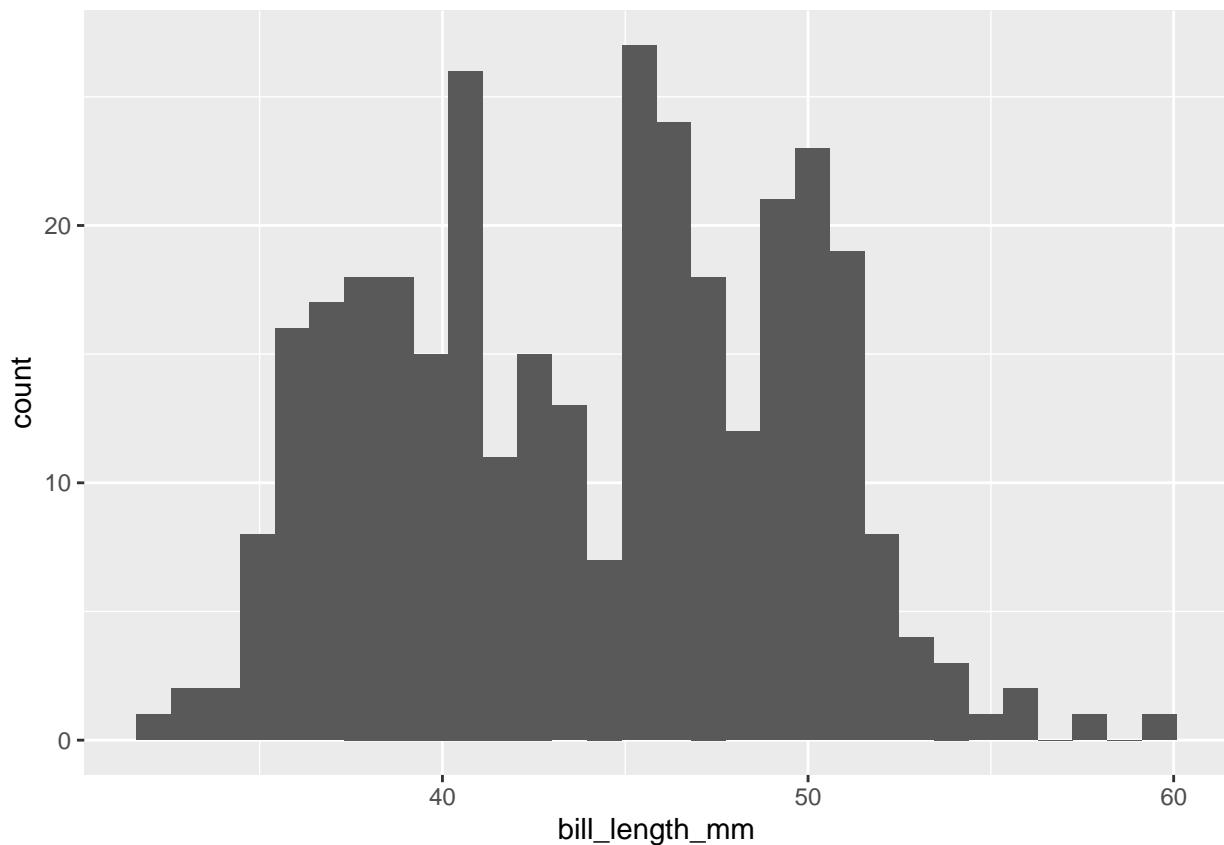
```
dplyr::glimpse(penguins)

## # Rows: 333
## # Columns: 20
## # $ x1          <dbl> 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 1~
## # $ study_name   <chr> "PAL0708", "PAL0708", "PAL0708", "PAL0708~
## # $ sample_number <dbl> 1, 2, 3, 5, 6, 7, 8, 13, 14, 15, 16, 17, 18, 19, 20,~
## # $ species       <chr> "Adelie Penguin (Pygoscelis adeliae)", "Adelie Penguin~
## # $ region        <chr> "Anvers", "Anvers", "Anvers", "Anvers", "Anvers", "A~
## # $ island         <chr> "Torgersen", "Torgersen", "Torgersen", "Torgersen", ~
## # $ stage          <chr> "Adult, 1 Egg Stage", "Adult, 1 Egg Stage", "Adult, ~
## # $ individual_id  <chr> "N1A1", "N1A2", "N2A1", "N3A1", "N3A2", "N4A1", "N4A~
## # $ clutch_completion <chr> "Yes", "Yes", "Yes", "Yes", "No", "No", "Yes"~
## # $ date_egg        <chr> "11/11/07", "11/11/07", "11/16/07", "11/16/07", "11/~
## # $ bill_length_mm   <dbl> 39.1, 39.5, 40.3, 36.7, 39.3, 38.9, 39.2, 41.1, 38.6~
## # $ bill_depth_mm    <dbl> 18.7, 17.4, 18.0, 19.3, 20.6, 17.8, 19.6, 17.6, 21.2~
## # $ flipper_length_mm <dbl> 181, 186, 195, 193, 190, 181, 195, 182, 191, 198, 18~
## # $ body_mass_g      <dbl> 3750, 3800, 3250, 3450, 3650, 3625, 4675, 3200, 3800~
## # $ sex              <chr> "MALE", "FEMALE", "FEMALE", "FEMALE", "MALE", "FEMAL~
## # $ delta_15_n_o_oo   <dbl> NA, 8.94956, 8.36821, 8.76651, 8.66496, 9.18718, 9.4~
## # $ delta_13_c_o_oo   <dbl> NA, -24.69454, -25.33302, -25.32426, -25.29805, -25.~
## # $ comments          <chr> "Not enough blood for isotopes.", NA, NA, NA, NA, "N~
## # $ spp               <chr> "Adelie", "Adelie", "Adelie", "Adelie", "Adelie", "A~
## # $ date_e            <date> 2007-11-11, 2007-11-11, 2007-11-16, 2007-11-16, 200~
```

```
# Every ggplot has 3 components: data, aes (aka mapping), geom:
ggplot(penguins, aes(x = bill_length_mm)) +
  geom_histogram()
```



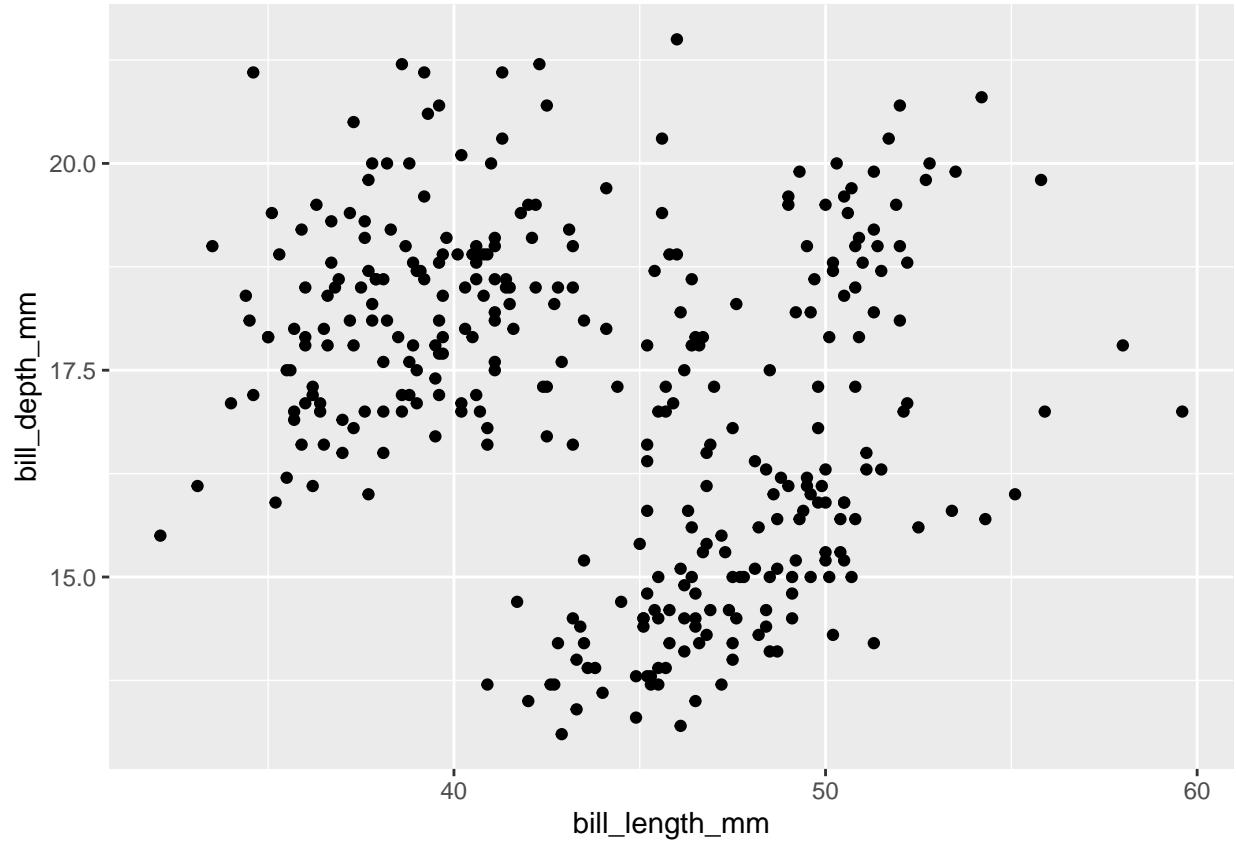
```
ggplot(penguins, aes(x = bill_length_mm)) +  
  geom_histogram()
```



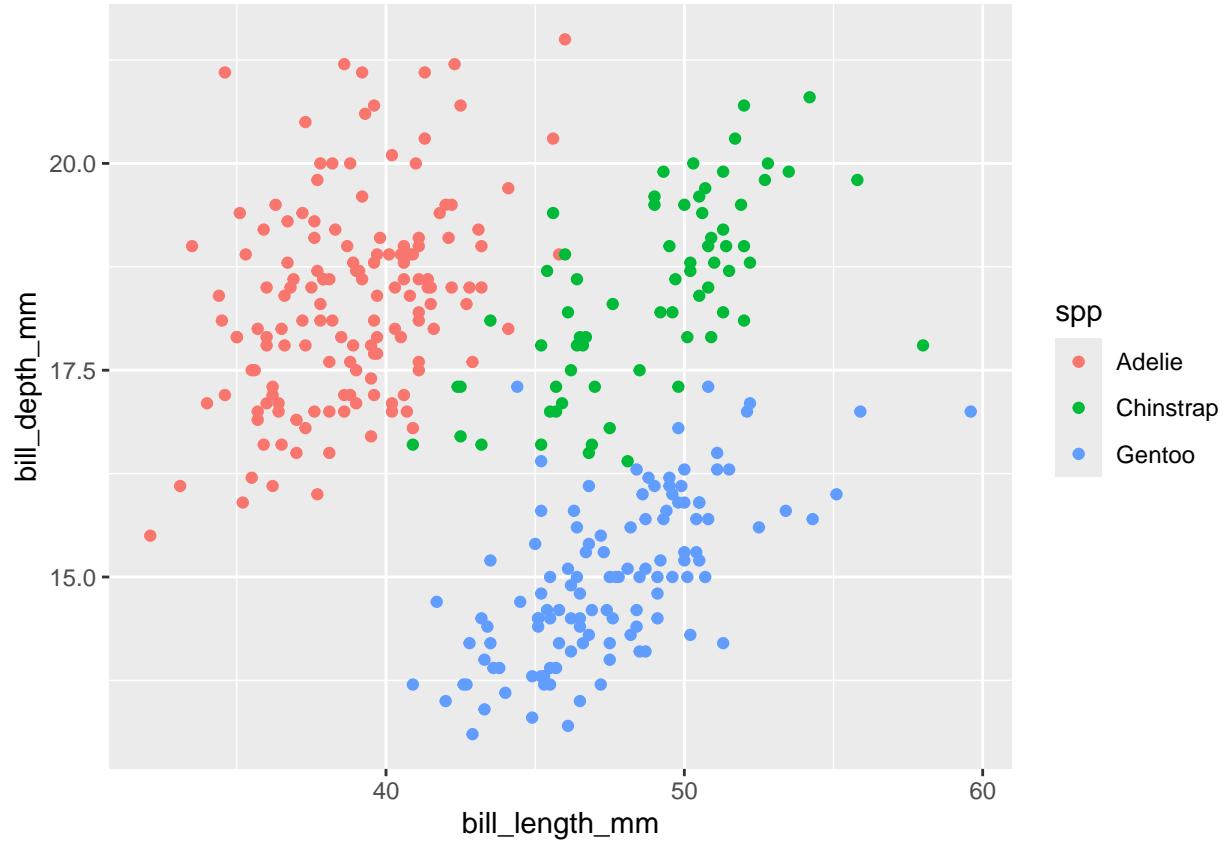
2. Plotting

2.1 Scatterplot

```
# BASE
ggplot(penguins, aes(x = bill_length_mm, y = bill_depth_mm)) +
  geom_point()
```



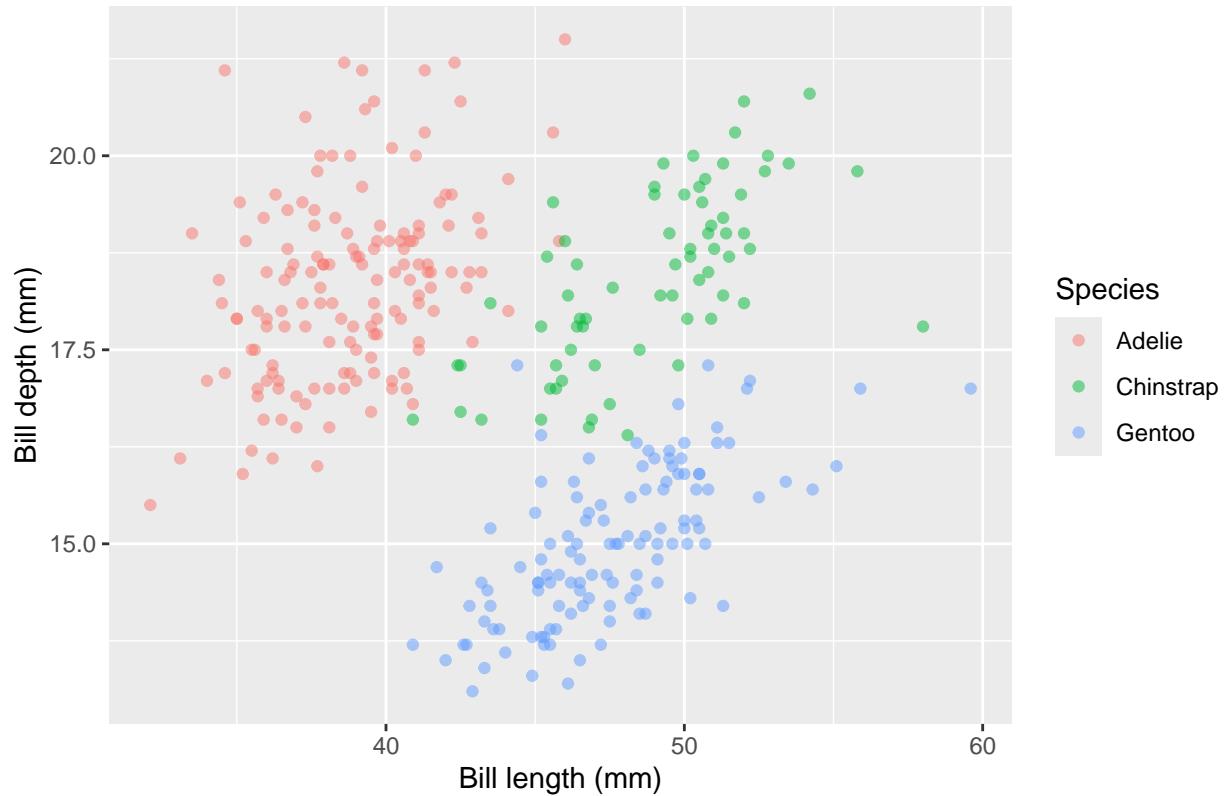
```
# Add color
ggplot(penguins, aes(x = bill_length_mm, y = bill_depth_mm, color = spp)) +
  geom_point()
```



```
# # Move legend
# ggplot(penguins, aes(x = bill_length_mm, y = bill_depth_mm, color = spp)) +
#   geom_point() +
#   theme(legend.position = "bottom")

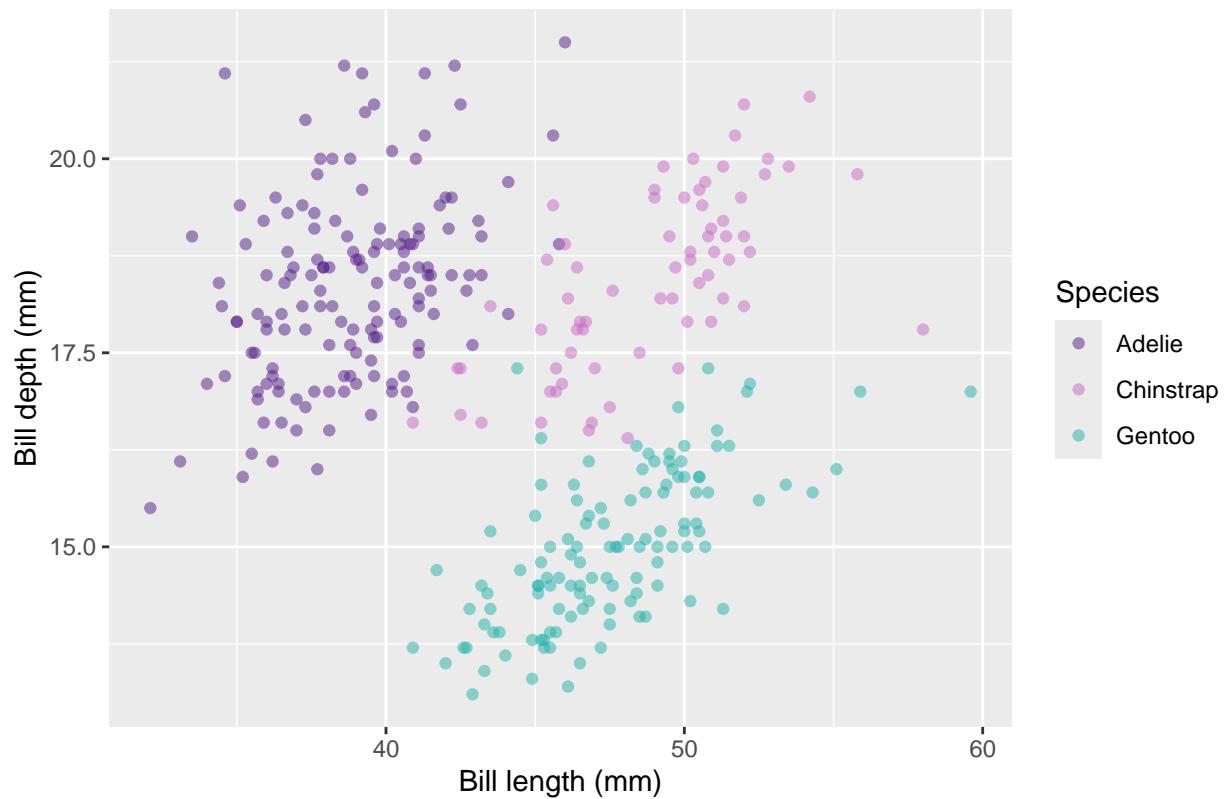
# Improve readability - alpha & labels & theme!
ggplot(penguins, aes(x = bill_length_mm, y = bill_depth_mm, color = spp)) +
  geom_point(alpha = 0.5) +
  labs(
    title = "Bill size varies by species",
    x = "Bill length (mm)",
    y = "Bill depth (mm)",
    color = "Species"
  )
```

Bill size varies by species



```
# Custom colors
ggplot(penguins, aes(x = bill_length_mm, y = bill_depth_mm, color = spp)) +
  geom_point(alpha = 0.5) +
  labs(
    title = "Bill size varies by species",
    x = "Bill length (mm)",
    y = "Bill depth (mm)",
    color = "Species"
  ) +
  scale_color_manual(values = c("purple4", "orchid3", "lightseagreen"))
```

Bill size varies by species



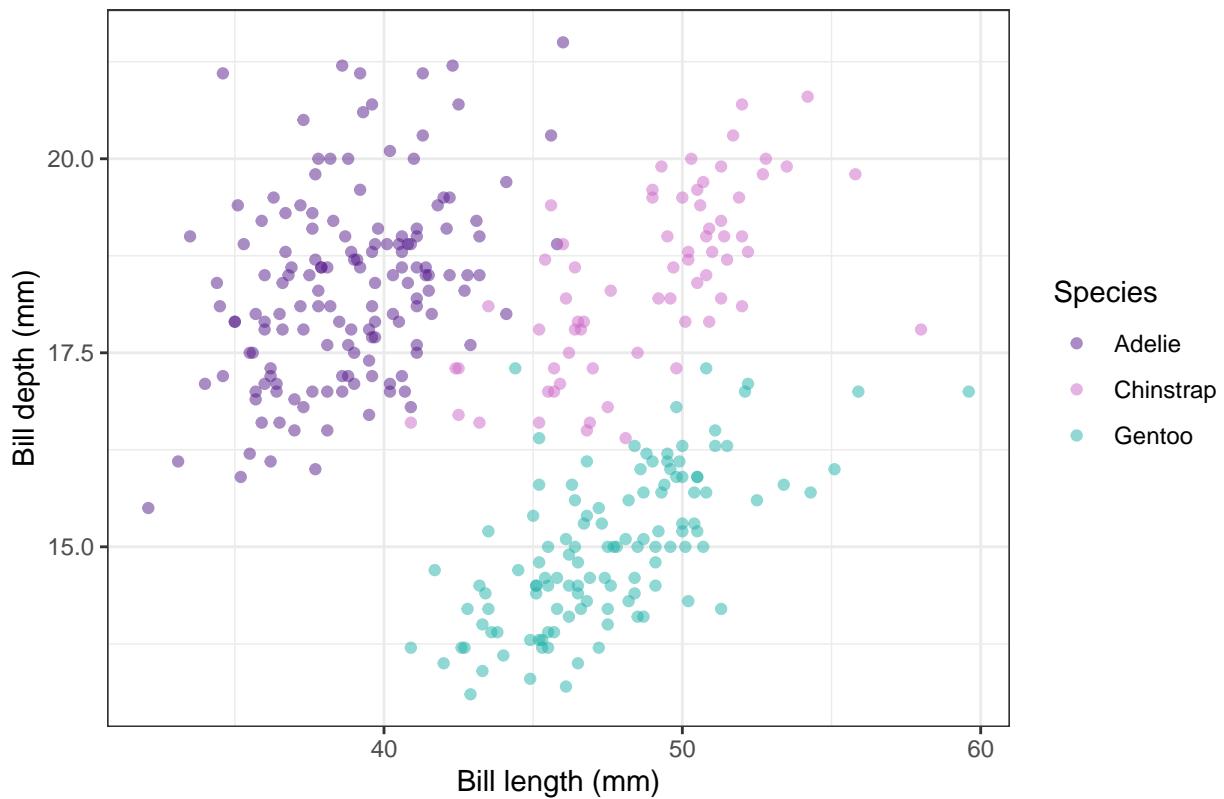
You can find all the keys and hex codes to innate R colors here: <https://r-charts.com/colors/>
... and more on how to use colors in ggplots here: <https://r-graph-gallery.com/ggplot2-color.html>

2.2 Adding themes

Themes are pre-defined sets of parameters that give ggplots specific types of appearances

```
ggplot(penguins, aes(x = bill_length_mm, y = bill_depth_mm, color = spp)) +
  geom_point(alpha = 0.5) +
  labs(
    title = "Bill size varies by species",
    x = "Bill length (mm)",
    y = "Bill depth (mm)",
    color = "Species"
  ) +
  scale_color_manual(values = c("purple4", "orchid3", "lightseagreen")) +
  theme_bw()
```

Bill size varies by species



```
# theme_light()
# theme_dark()
# theme_minimal()
# theme_classic() # classic data plot ready for publication
# theme_void() #excludes anything but the geom itself

# you can save the plot and define the dimensions!
ggsave("Bill size by species.png") # if you don't give dimensions, it'll appear like it does in your Pl
ggsave("Bill size by species 5x4.png", width = 5, height = 4) # but you can also define dimensions.
# do this to keep multiple plot dimensions consistent in your publication/poster!
```

More on themes here: <https://ggplot2.tidyverse.org/reference/ggtheme.html>

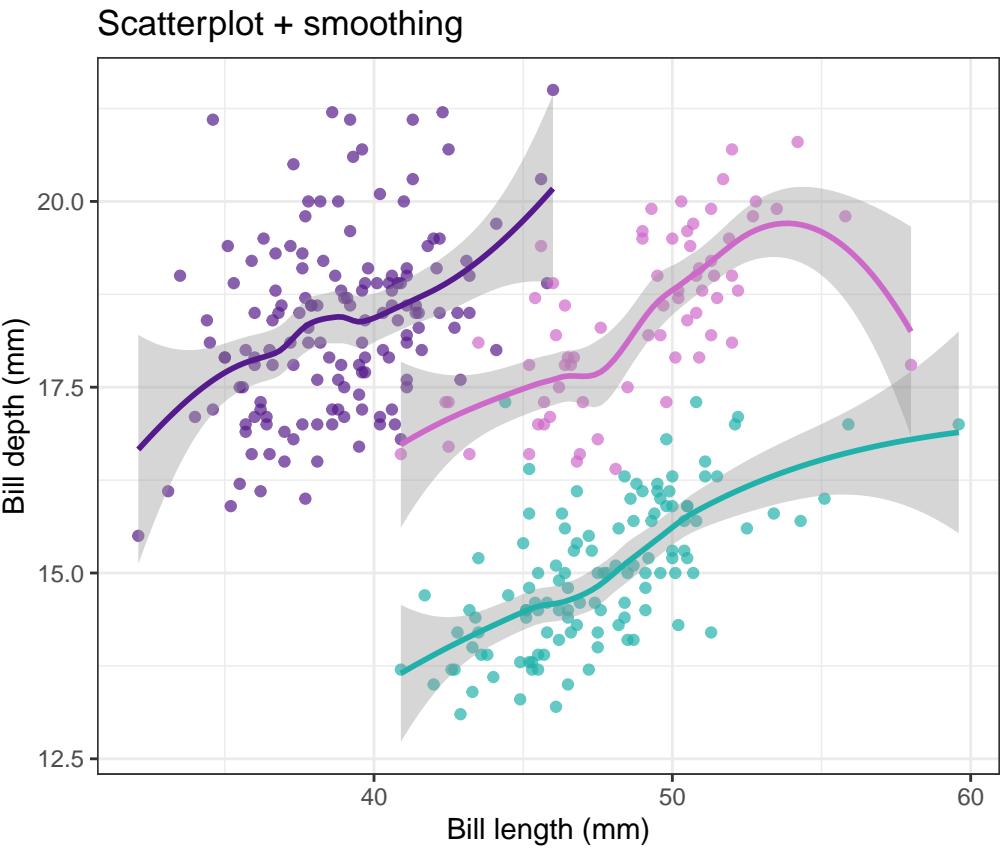
2.3 Add trendlines

```
# smooth (curved) lines
ggplot(penguins, aes(x = bill_length_mm, y = bill_depth_mm, color = spp)) +
  geom_point(alpha = 0.7) +
  geom_smooth() +
  labs(
    title = "Scatterplot + smoothing",
    x = "Bill length (mm)",
```

```

y = "Bill depth (mm)",
color = "Species"
) +
scale_color_manual(values = c("purple4", "orchid3", "lightseagreen")) +
theme_bw()

```

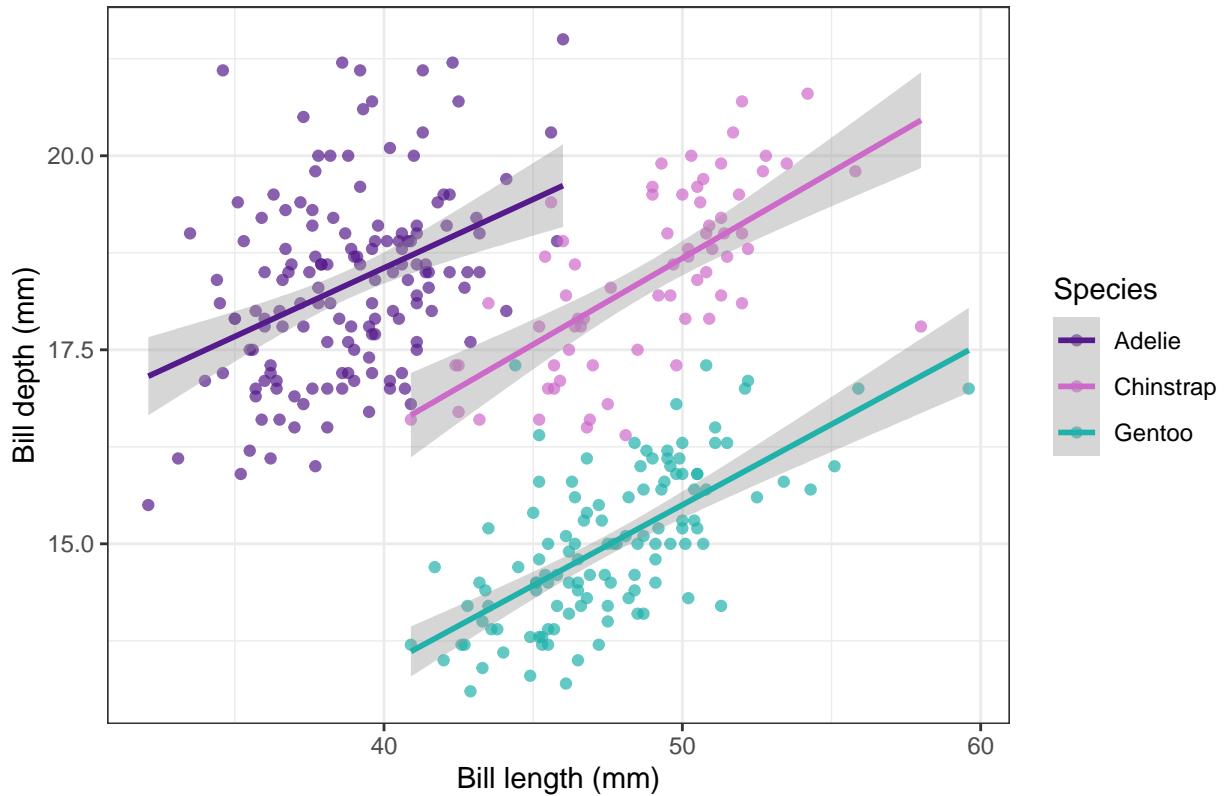


```

# make it linear!
ggplot(penguins, aes(x = bill_length_mm, y = bill_depth_mm, color = spp)) +
  geom_point(alpha = 0.7) +
  geom_smooth(method = "lm") +
  labs(
    title = "Scatterplot + smoothing",
    x = "Bill length (mm)",
    y = "Bill depth (mm)",
    color = "Species"
  ) +
  scale_color_manual(values = c("purple4", "orchid3", "lightseagreen")) +
  theme_bw()

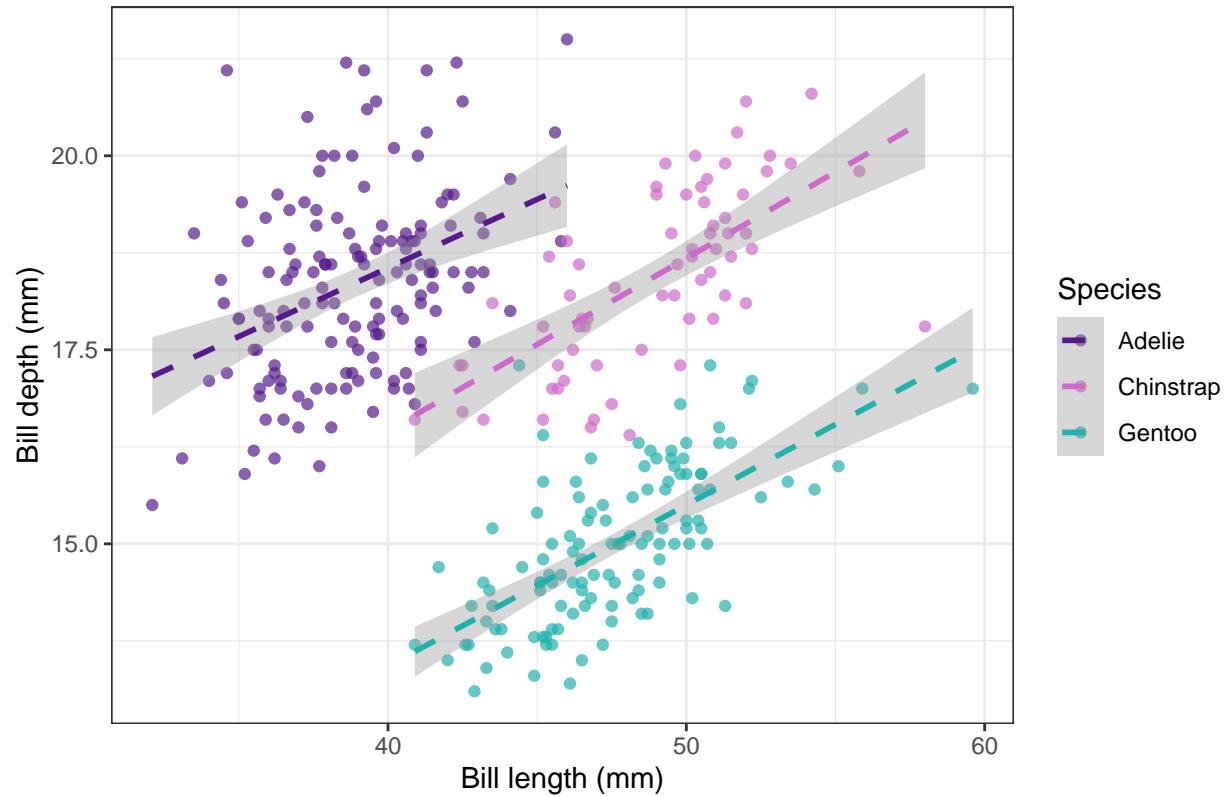
```

Scatterplot + smoothing



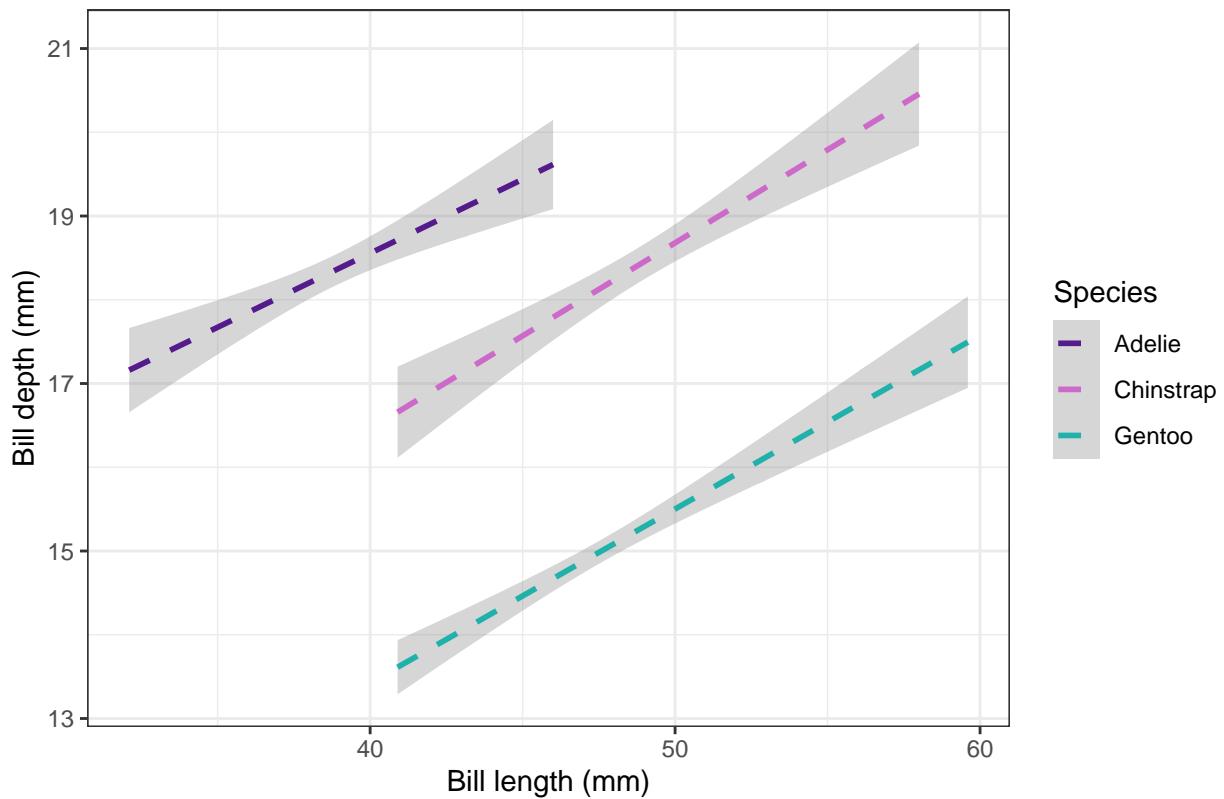
```
# Modify the lines
ggplot(penguins, aes(x = bill_length_mm, y = bill_depth_mm, color = spp)) +
  geom_point(alpha = 0.7) +
  geom_smooth(method = "lm", linetype = "dashed") +
  labs(
    title = "Scatterplot + smoothing",
    x = "Bill length (mm)",
    y = "Bill depth (mm)",
    color = "Species"
  ) +
  scale_color_manual(values = c("purple4", "orchid3", "lightseagreen")) +
  theme_bw()
```

Scatterplot + smoothing



```
# or only have lines:  
ggplot(penguins, aes(x = bill_length_mm, y = bill_depth_mm, color = spp)) +  
  geom_smooth(method = "lm", linetype = "dashed") +  
  labs(  
    title = "Linear Regression",  
    x = "Bill length (mm)",  
    y = "Bill depth (mm)",  
    color = "Species"  
) +  
  scale_color_manual(values = c("purple4", "orchid3", "lightseagreen")) +  
  theme_bw()
```

Linear Regression

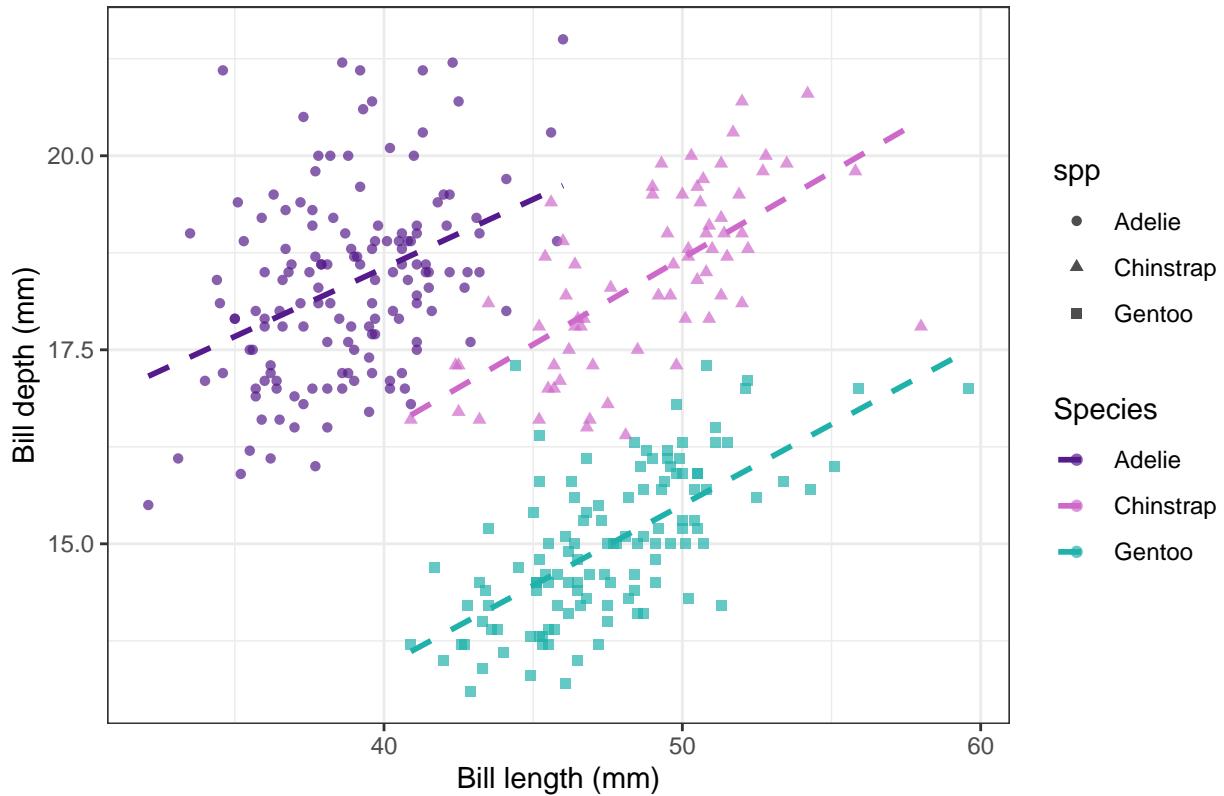


Other linetype options are: twodash; solid; longdash; dotted; dotdash; dashed.

2.4 Shapes

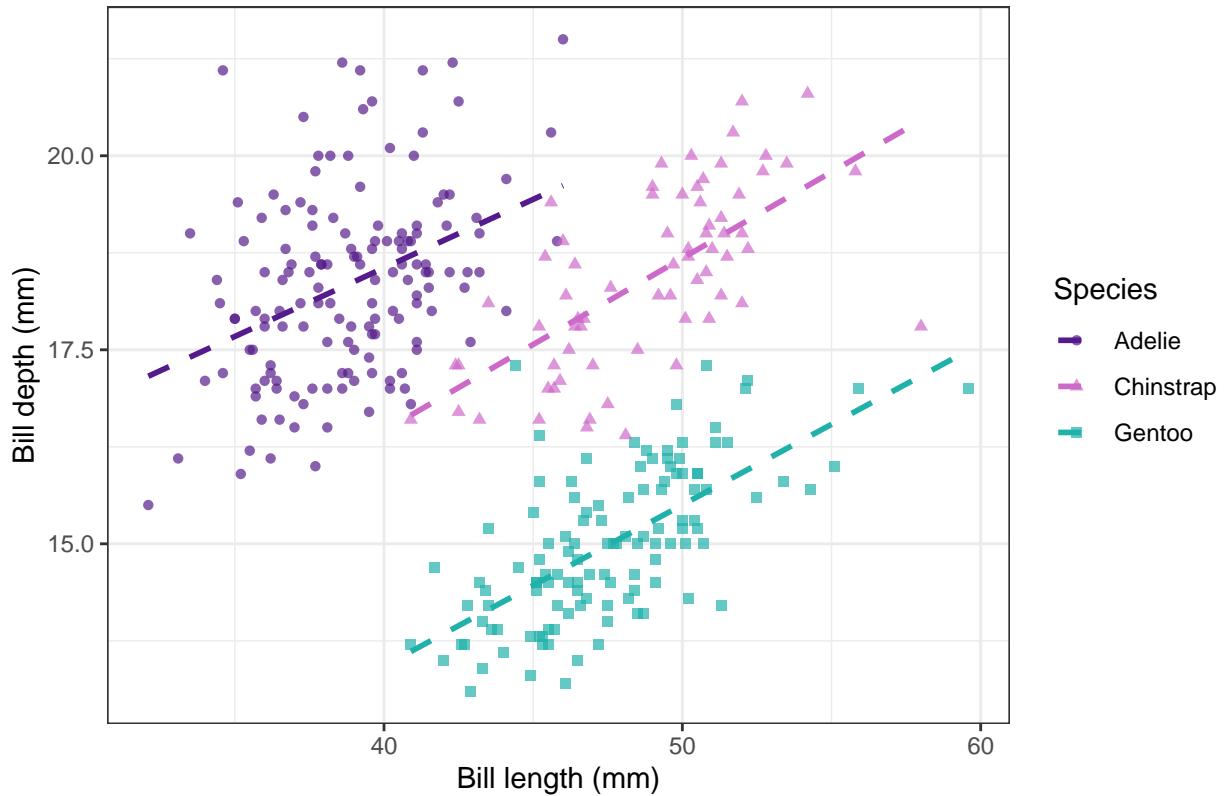
```
ggplot(penguins, aes(x = bill_length_mm, y = bill_depth_mm, color = spp, shape = spp)) +  
  geom_point(alpha = 0.7) +  
  geom_smooth(method = "lm", se = FALSE, linetype = "dashed") +  
  labs(  
    title = "Scatterplot + smoothing",  
    x = "Bill length (mm)",  
    y = "Bill depth (mm)",  
    color = "Species"  
) +  
  scale_color_manual(values = c("purple4", "orchid3", "lightseagreen")) +  
  theme_bw()
```

Scatterplot + smoothing



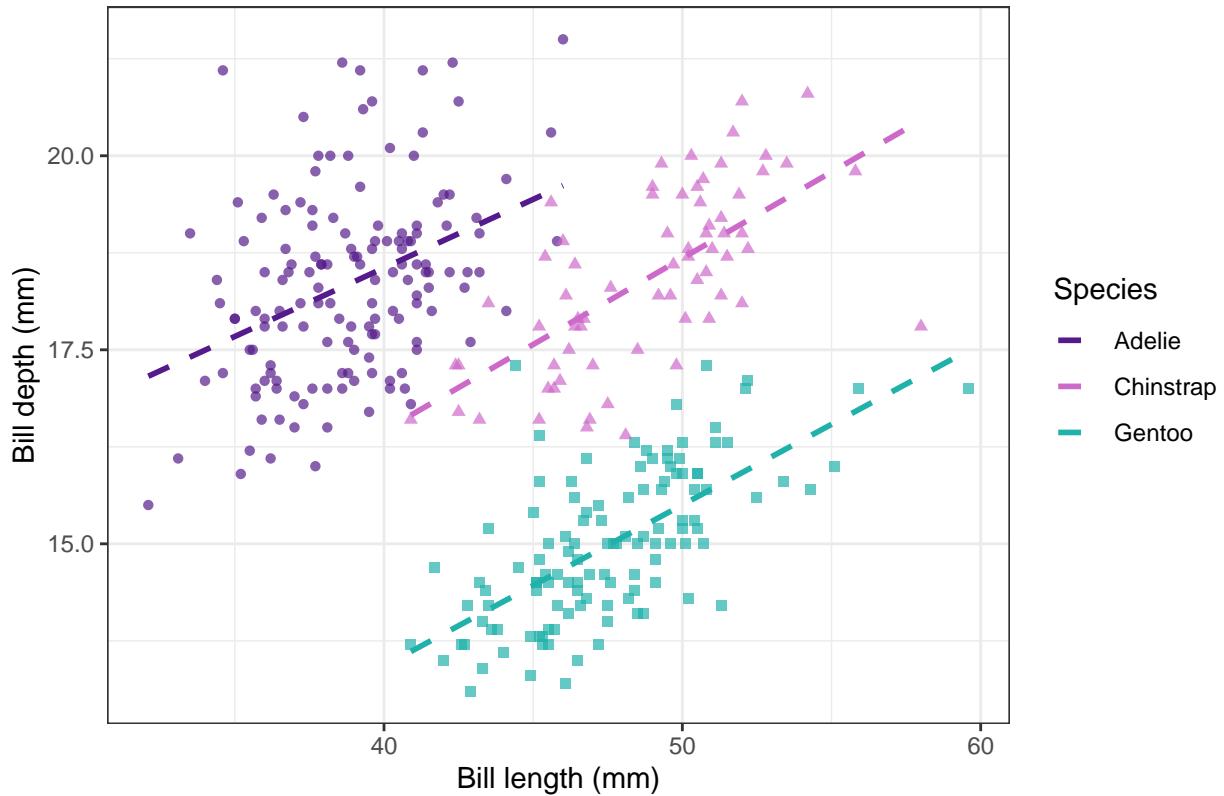
```
# combine the legends
ggplot(penguins, aes(x = bill_length_mm, y = bill_depth_mm, color = spp, shape = spp)) +
  geom_point(alpha = 0.7) +
  geom_smooth(method = "lm", se = FALSE, linetype = "dashed") +
  labs(
    title = "Scatterplot + smoothing",
    x = "Bill length (mm)",
    y = "Bill depth (mm)",
    color = "Species",
    shape = "Species" # ----- note that we tell ggplot that the legend for the shape has the same name
  ) +
  scale_color_manual(values = c("purple4", "orchid3", "lightseagreen")) +
  theme_bw()
```

Scatterplot + smoothing



```
# ...or...
ggplot(penguins, aes(x = bill_length_mm, y = bill_depth_mm, color = spp, shape = spp)) +
  geom_point(alpha = 0.7, show.legend = FALSE) + # < ----- just tell geom_point NOT to plot legend at all
  geom_smooth(method = "lm", se = FALSE, linetype = "dashed") +
  labs(
    title = "Scatterplot + smoothing",
    x = "Bill length (mm)",
    y = "Bill depth (mm)",
    color = "Species"
    # no "shape" in labels section
  ) +
  scale_color_manual(values = c("purple4", "orchid3", "lightseagreen")) +
  theme_bw()
```

Scatterplot + smoothing

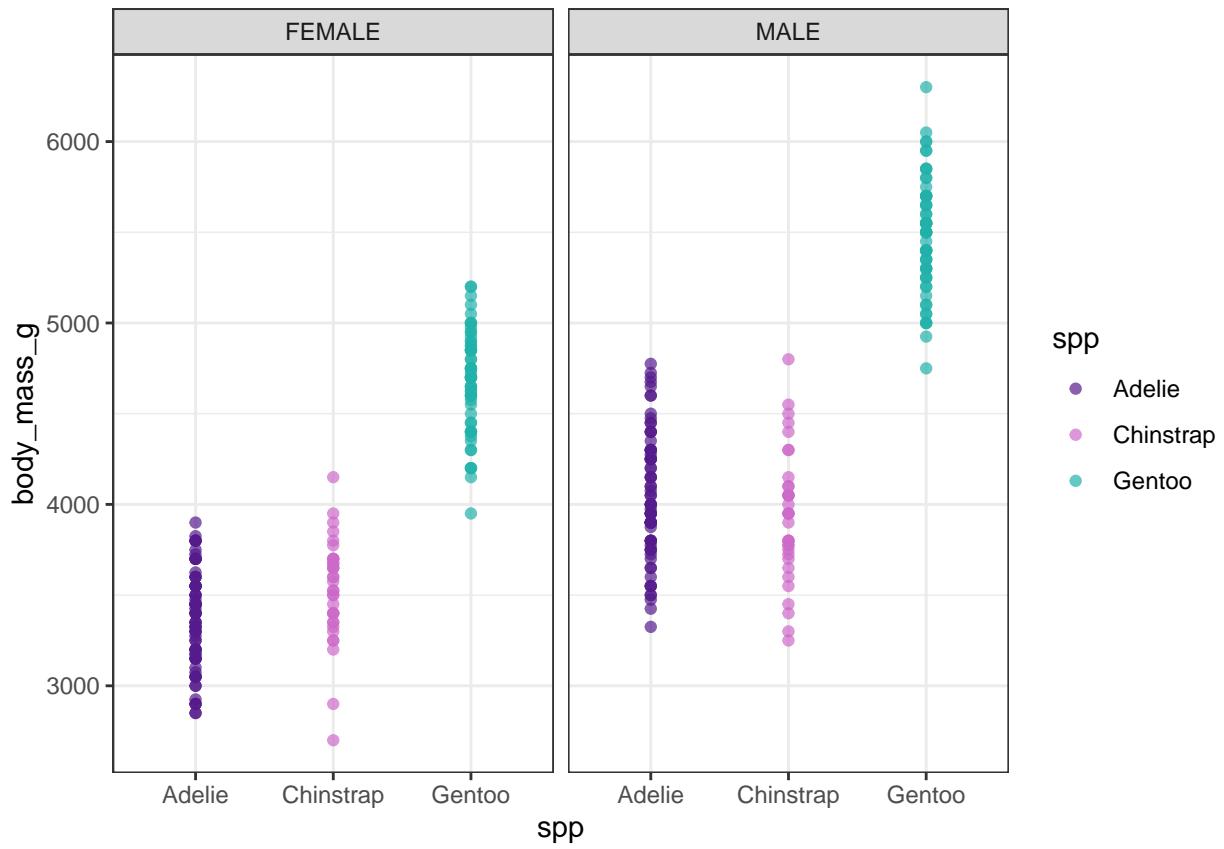


```
# Now ggplot knows color and shape belong to the same variable. It automatically merges them into a si
```

2.5 Facetting

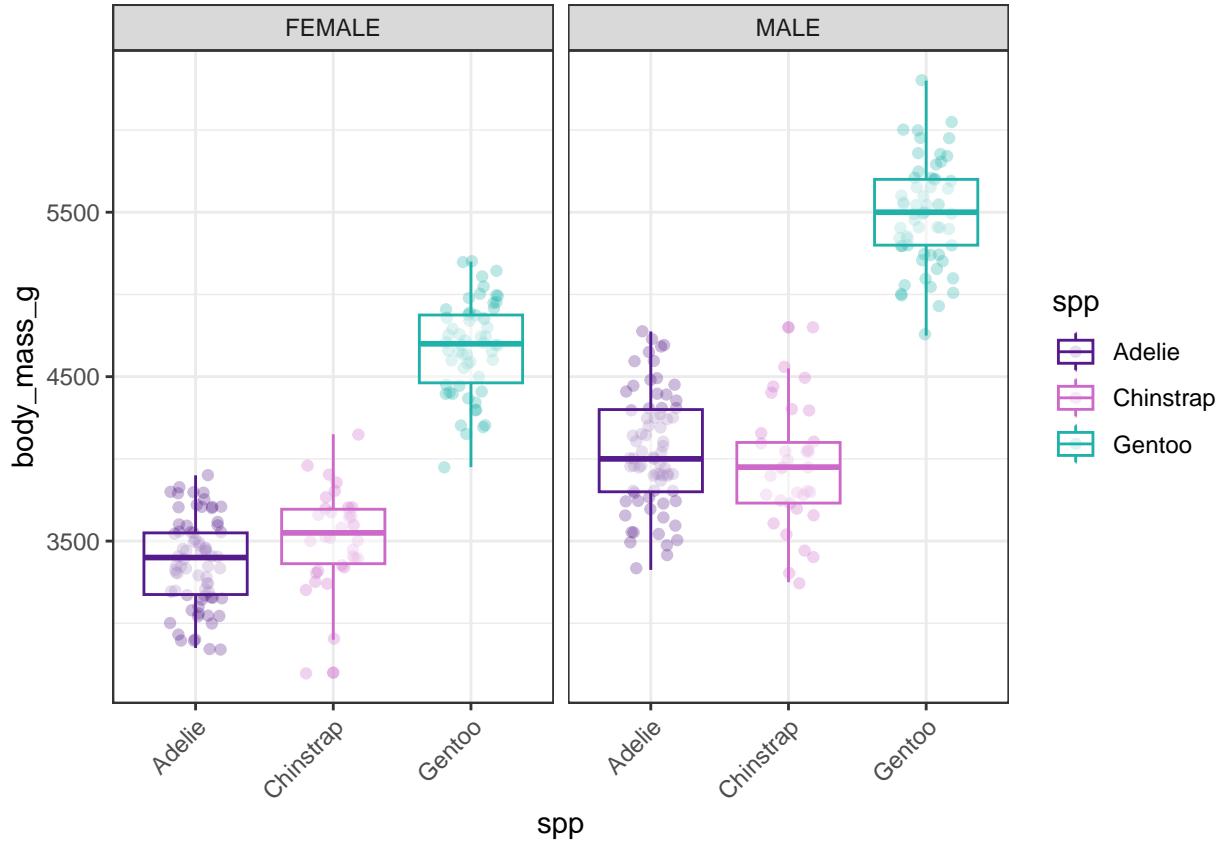
Also, boxplots and axis text

```
# you can divide by a categorical variable
ggplot(penguins, aes(x = spp, y = body_mass_g, color = spp)) +
  geom_point(alpha = 0.7) +
  facet_wrap(~ sex) +
  scale_color_manual(values = c("purple4", "orchid3", "lightseagreen")) +
  theme_bw()
```



```
# You can use this to show relationships between three variables: species, sex, body mass

# Boxplots might be a better geom here:
ggplot(penguins, aes(x = spp, y = body_mass_g, color = spp)) +
  geom_jitter(alpha = 0.3, width = 0.2) +
  geom_boxplot(alpha = 0.5) +
  facet_wrap(~ sex) +
  scale_color_manual(values = c("purple4", "orchid3", "lightseagreen")) +
  theme_bw() +
  theme(axis.text.x = element_text(angle = 45, vjust = 1, hjust = 1))
```



To use `vjust` and `hjust` for axis text in ggplot2, modify the `theme()` function with `axis.text.x` or `axis.text.y` and the `element_text()` function. The values for `vjust` (vertical justification) and `hjust` (horizontal justification) range from 0 to 1.

Next, we'll look at how to manage your viewer's attention, and in the process, we'll see new types of charts.

3. Storytelling: Managing the viewer's attention

3.1 Using colors

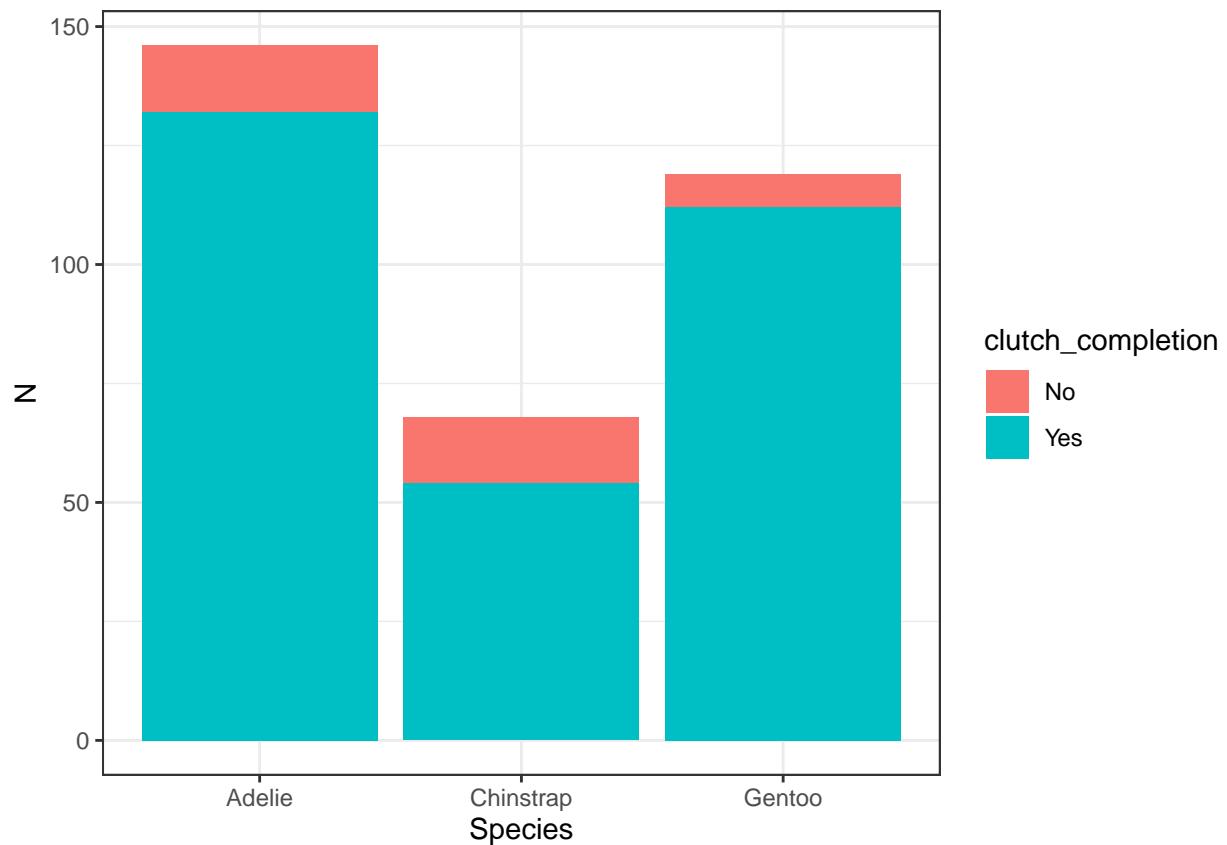
```
# let's look at some other aspects of the data
penguins %>% dplyr::count(clutch_completion, spp)
```

```
## # A tibble: 6 x 3
##   clutch_completion spp      n
##   <chr>            <chr>    <int>
## 1 No               Adelie     14
## 2 No               Chinstrap   14
## 3 No               Gentoo     7
## 4 Yes              Adelie    132
## 5 Yes              Chinstrap  54
## 6 Yes              Gentoo    112
```

```

# basic stacked bar chart
ggplot(penguins, aes(x = spp, fill = clutch_completion)) +
  geom_bar() +
  # scale_fill_manual(values = c("green4", "firebrick3")) +
  # but notice the colors, what are they telling you???
  # scale_fill_manual(values = c("firebrick", "green4")) +
  # this makes more sense-- play on peoples' familiarity with color-coding!!!
  labs(x = "Species", y = "N") +
  theme_bw()

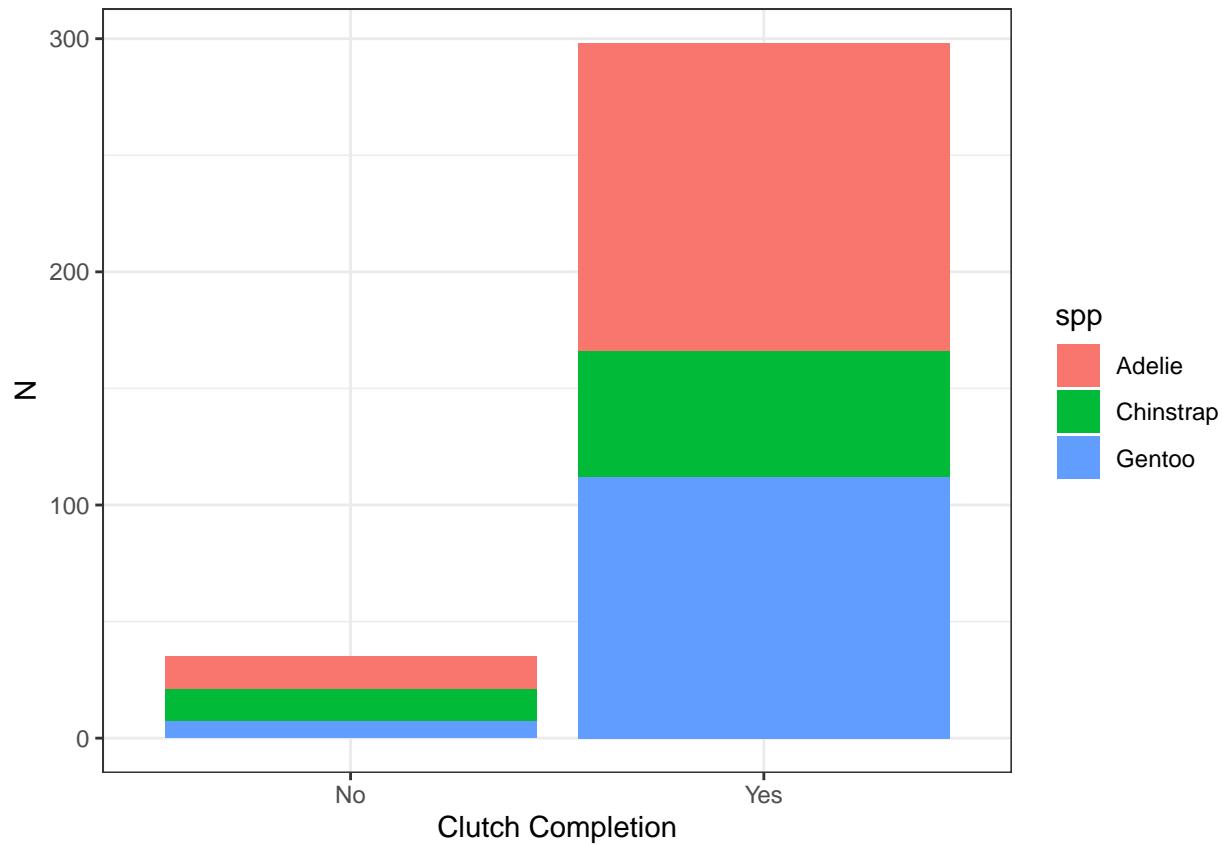
```



```

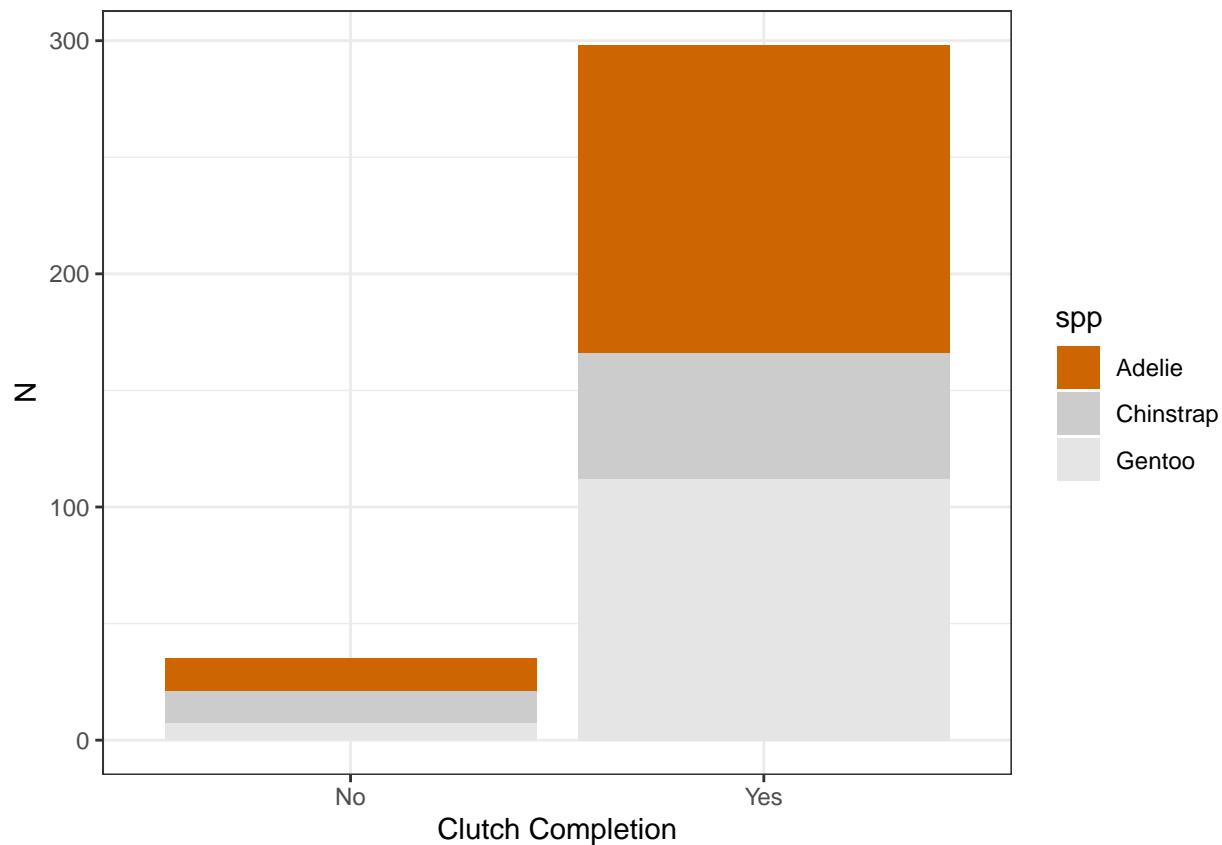
# you can change the focus of the chart (swap color vs. x-axis)
ggplot(penguins, aes(x = clutch_completion, fill = spp)) +
  geom_bar() +
  #scale_fill_manual(values = c("purple4", "orchid3", "lightseagreen")) +
  labs(x = "Clutch Completion", y = "N") +
  theme_bw()

```



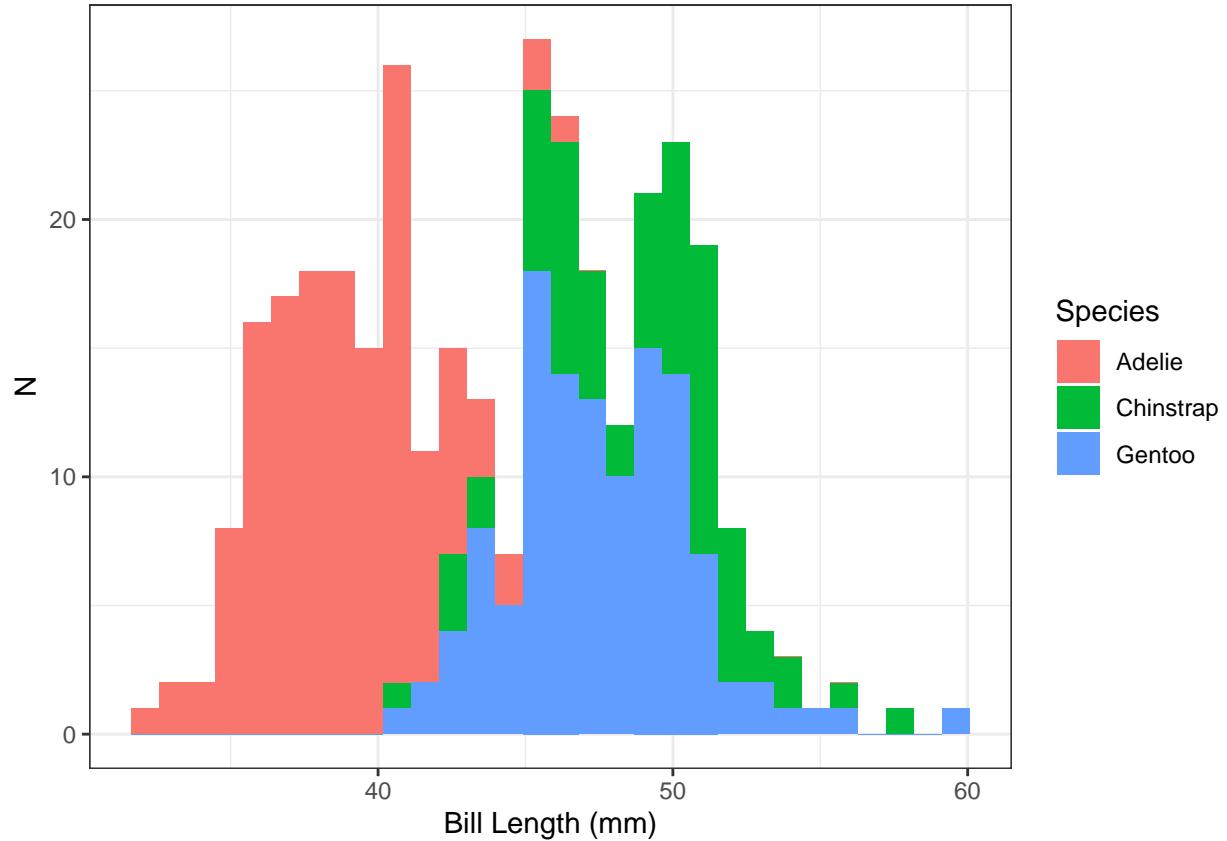
```
# --> we simply switched which variable was on the x-axis, and while is the color fill, but it makes a
```

```
# To draw the viewer's attention, you can also highlight one particular aspect using colors
ggplot(penguins, aes(x = clutch_completion, fill = spp)) +
  geom_bar() +
  scale_fill_manual(values = c("darkorange3", "grey80", "grey90")) +
  labs(x = "Clutch Completion", y = "N") +
  theme_bw()
```

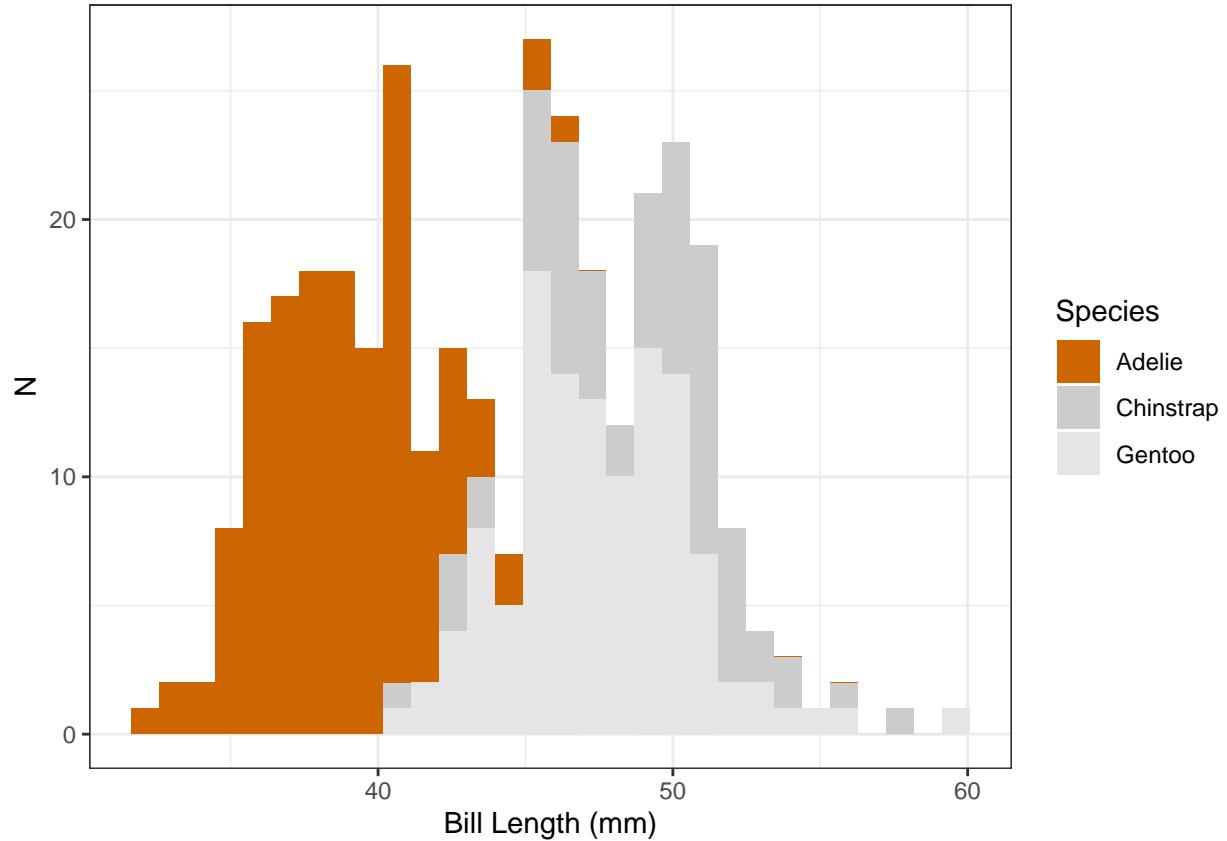


```
#####
# ADD N = numbers to bars???
#####

# you can also go back to your summary stats (the histogram):
ggplot(penguins, aes(x = bill_length_mm, fill = spp)) +
  geom_histogram() +
  labs(x = "Bill Length (mm)", y = "N", fill = "Species") +
  theme_bw()
```



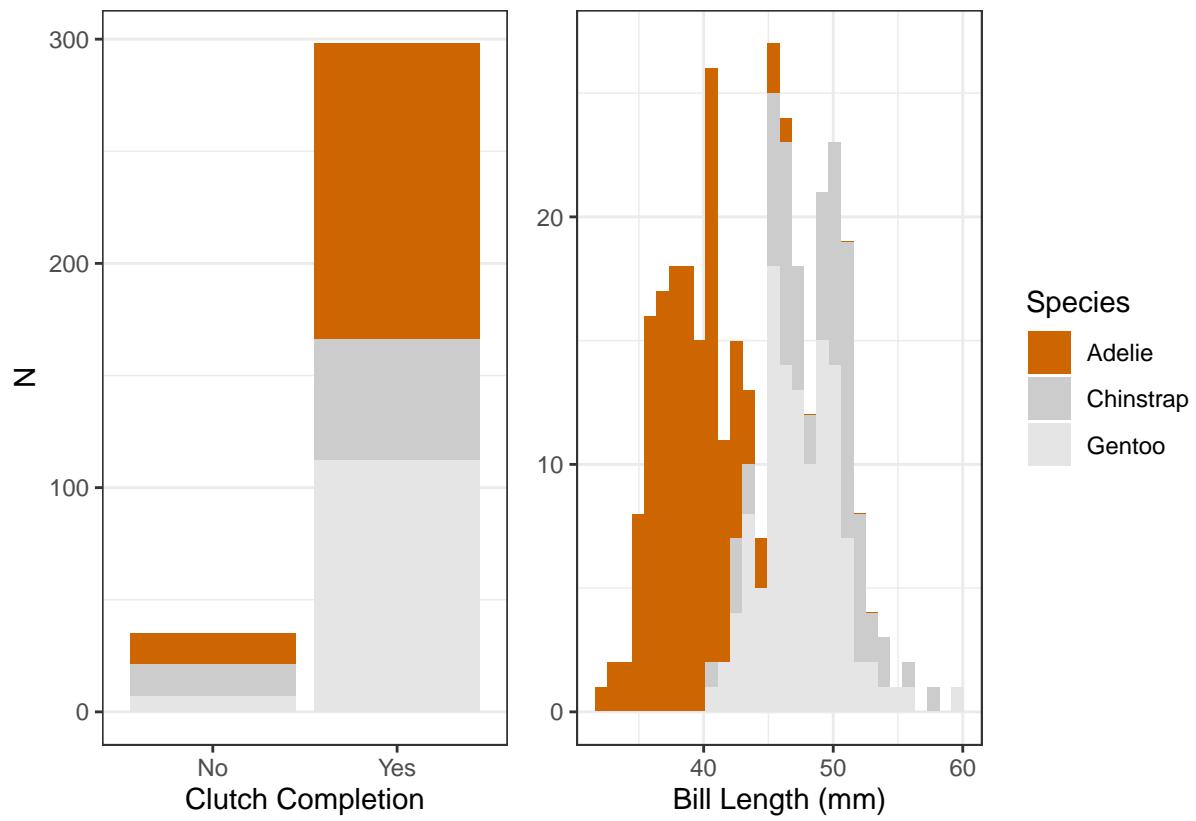
```
ggplot(penguins, aes(x = bill_length_mm, fill = spp)) +
  geom_histogram() +
  scale_fill_manual(values = c("darkorange3", "grey80", "grey90")) +
  labs(x = "Bill Length (mm)", y = "N", fill = "Species") +
  theme_bw()
```



```
# -----
plot1 <- ggplot(penguins, aes(x = clutch_completion, fill = spp)) +
  geom_bar() +
  scale_fill_manual(values = c("darkorange3", "grey80", "grey90")) +
  labs(x = "Clutch Completion", y = "N", fill = "Species") +
  theme_bw() +
  theme(panel.grid.major.x = element_blank())

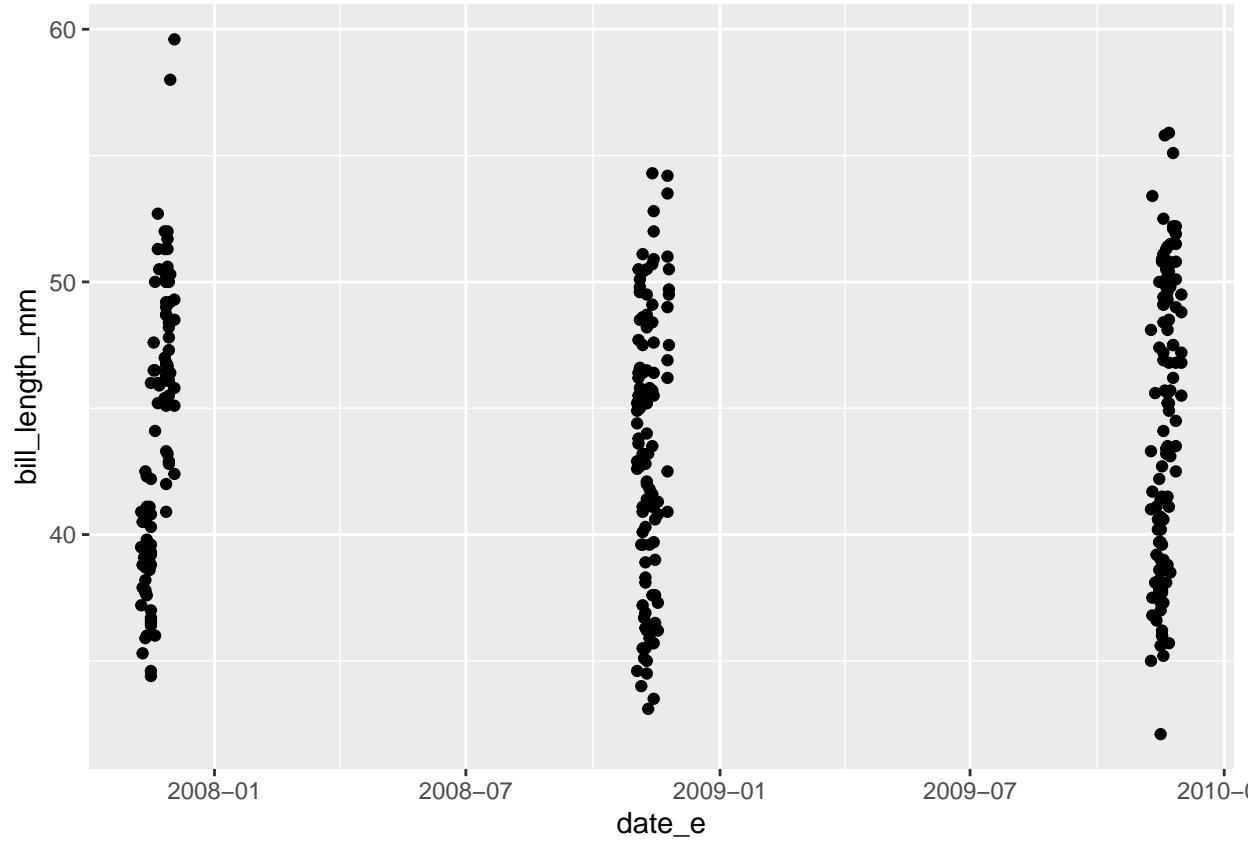
plot2 <- ggplot(penguins, aes(x = bill_length_mm, fill = spp)) +
  geom_histogram() +
  scale_fill_manual(values = c("darkorange3", "grey80", "grey90")) +
  labs(x = "Bill Length (mm)", y = "N", fill = "Species") +
  theme_bw()

# this is part of the package {patchwork}
plot1 + plot2 + plot_layout(axes = "collect", guides = "collect")
```



3.2 Using visual cues

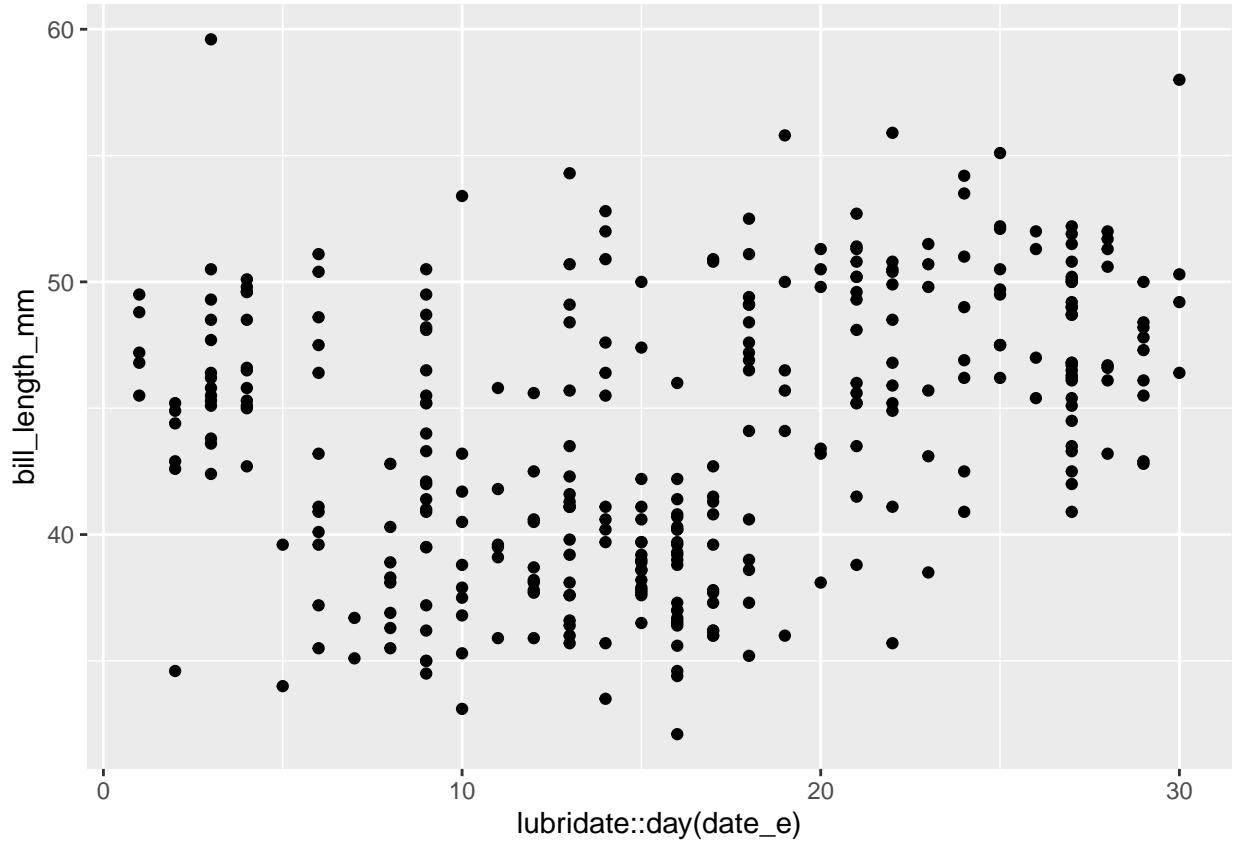
```
# let's look at bill length by egg hatching date 'date_e'
ggplot(penguins, aes(x = date_e, y = bill_length_mm)) +
  geom_point()
```



```
# hmm... looks like there was one season each year from 2007-2009...
# so, does it make sense to plot the x-axis as full dates? with months and days?
# seems like a waste of space.

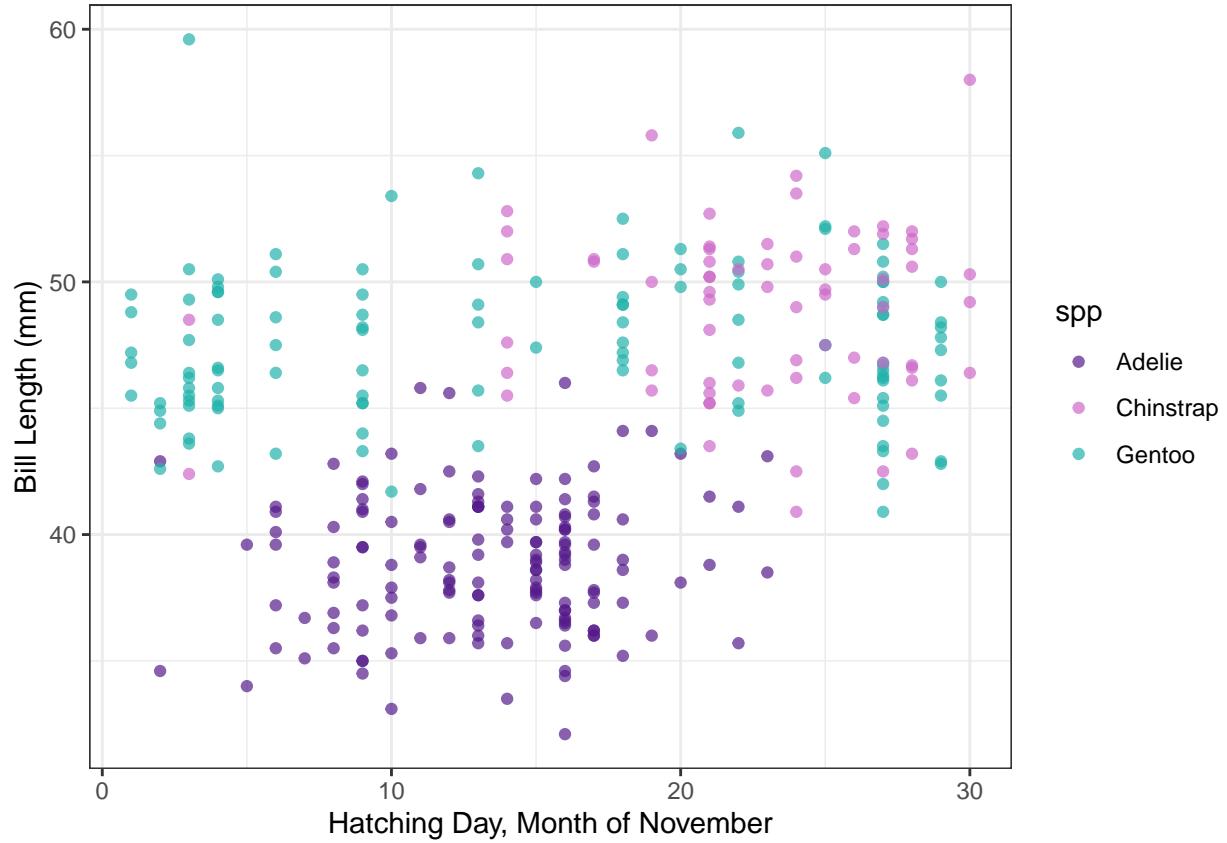
# instead, we can focus on what we want to look at-- egg-laying season, which seems to be days within the month
```

ggplot(penguins, aes(x = lubridate::day(date_e), y = bill_length_mm)) +
 geom_point()

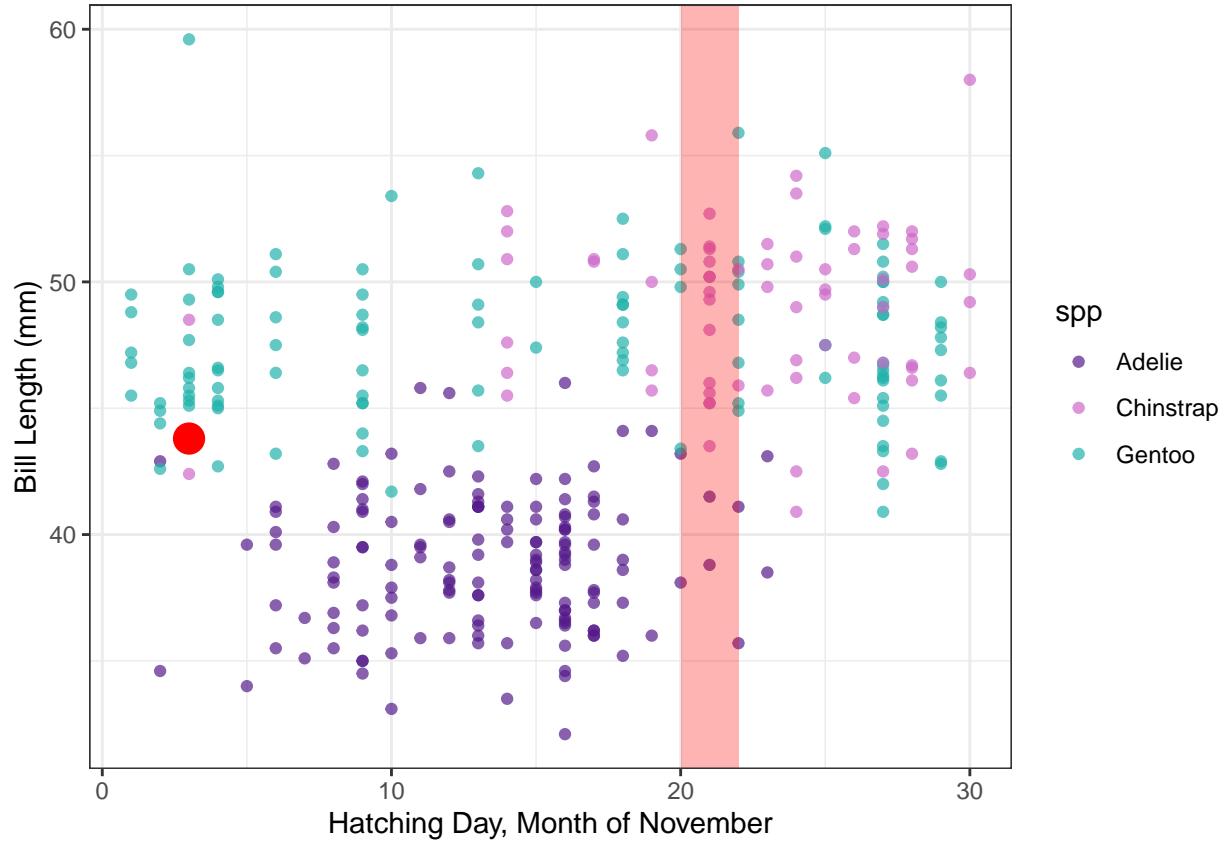


```
# good, but not there yet...

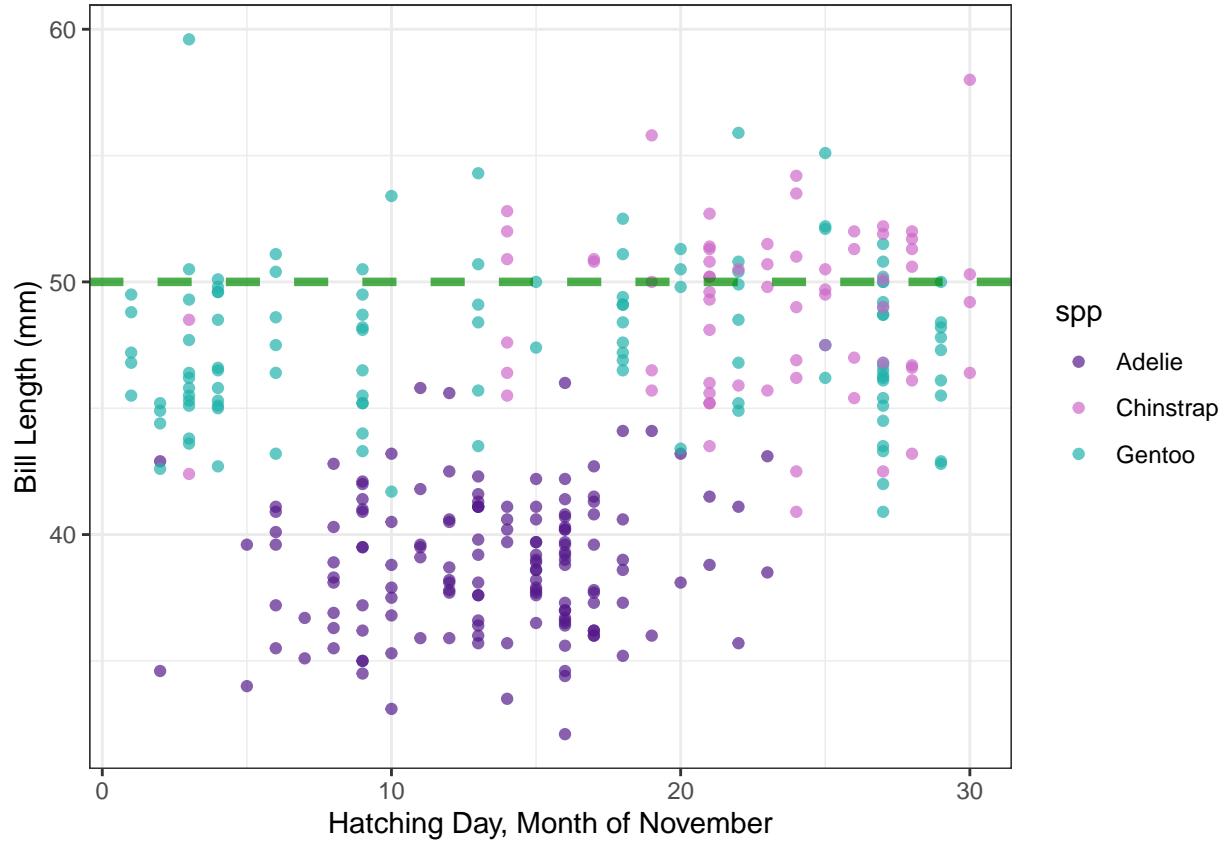
# let's try to see what's happening by species:
# add 'color = spp' to aes() in ggplot
ggplot(penguins, aes(x = lubridate::day(date_e), y = bill_length_mm, color = spp)) +
  geom_point(alpha = 0.7) +
  scale_color_manual(values = c("purple4", "orchid3", "lightseagreen")) +
  labs(x = "Hatching Day, Month of November", y = "Bill Length (mm)") +
  # scale_x_continuous(breaks = c(1:30)) +
  theme_bw()
```



```
# let's say something happened on a particular date that you want to highlight:
# for example, if a few days had rain:
ggplot(penguins, aes(x = lubridate::day(date_e), y = bill_length_mm, color = spp)) +
  geom_point(alpha = 0.7) +
  scale_color_manual(values = c("purple4", "orchid3", "lightseagreen")) +
  annotate("rect",
    xmin = 20, # from the 20th
    xmax = 22, # to the 22nd
    ymin = Inf, # set ymin to +infinity
    ymax = -Inf, # set ymax to -infinity
    fill = "red",
    # fill = "skyblue", # maybe use a color that makes sense for rain instead of red
    alpha = 0.3) +
  geom_point(data = penguins %>% filter(individual_id == "N16A1"), color = "red", size = 5) +
  # annotate("text", label = "<-- Monsoon", x = 25, y = 58, color = "blue") + # include text annotation
  labs(x = "Hatching Day, Month of November", y = "Bill Length (mm)") +
  theme_bw()
```



```
# or if you want to highlight a threshold, use a single line (instead of a rectangle):
ggplot(penguins, aes(x = lubridate::day(date_e), y = bill_length_mm, color = spp)) +
  geom_point(alpha = 0.7) +
  scale_color_manual(values = c("purple4", "orchid3", "lightseagreen")) +
  geom_hline(yintercept = 50, color = "green4", linetype = "dashed", linewidth = 1.5, alpha = 0.7) +
  # geom_vline(xintercept = 15, color = "red", linetype = "solid", linewidth = 1.5, alpha = 0.7) +
  labs(x = "Hatching Day, Month of November", y = "Bill Length (mm)") +
  theme_bw()
```



4. Primate Data

4.1 Explore

```
# let's have a look at the data:
glimpse(crossings)
```

```
## Rows: 791
## Columns: 13
## $ date
## $ crossing_id_number
## $ start_time
## $ group
## $ time_of_crossing
## $ id
## $ age_sex_class
## $ cross_gait
## $ mid_cross_behavior
## $ cross_complete
## $ post_cross_food_behavior
## $ post_cross_other_behavior
## $ notes
<chr> "7/29/25", "7/29/25", "7/29/25", "7/29/25", ~
<dbl> 10101, 10102, 10103, 10104, 10105, 10106, 10~
<time> 11:30:00, 11:30:00, 11:30:00, 11:30:00, 11:~
<chr> "B", "B", "B", "B", "B", "B", "B", "B", "B", ~
<chr> "11:32:00", "11:32:37", "11:32:37", "11:32:37", ~
<chr> "U", "U", "U", "Rimba", "U", "U", "Rimba", "~
<chr> "J", "J", "J", "SM", "J", "J", "SM", "J", "S~
<chr> "RN", "WK", "WK", "BWK", "RN", "WK", "WK", "~
<chr> "TC", NA, "TC", NA, NA, NA, "MST", FE", NA, N~
<chr> "C", "C", "C", "C", "C", "C", "C", "C", "C", ~
<chr> "FO", FE", "FE", NA, NA, NA, "FE", NA, "FO", ~
<chr> NA, NA, "MO", "SB", "PL", NA, "SG", NA, "SG"~
<chr> NA, ~
```

```

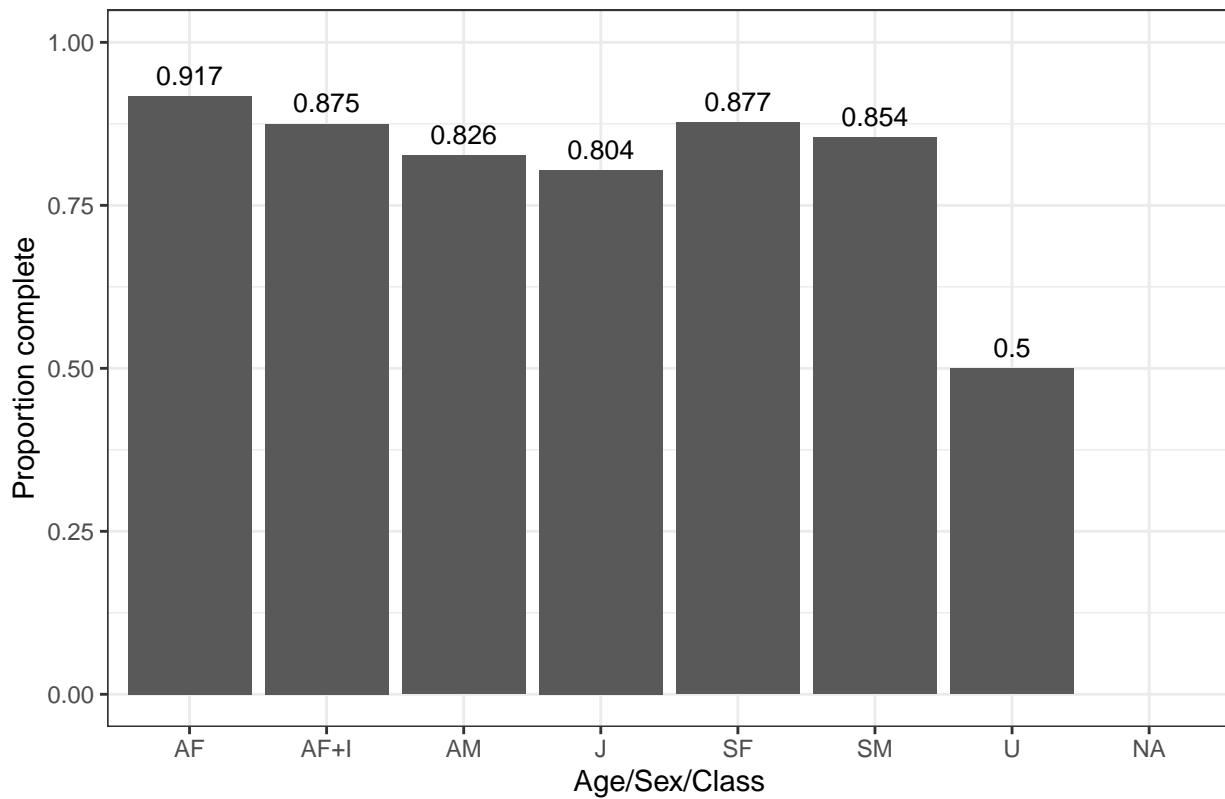
# Ok, no numeric variables. So, I'd want to compile them somehow.
# Let's group by a/s class, and count the number of crossings per a/s class
# and then convert that to a proportion
# Completion rate by sex
cross1 <- crossings %>%
  # dplyr::filter(!is.na(age_sex_class)) %>%
  group_by(age_sex_class) %>%
  summarise(
    n_crossings = n(),
    prop_complete = mean(cross_complete == "C", na.rm = TRUE)
  )

# This will give:
# Count per age class
# Proportion complete (numeric 0-1)

# plot:
# Bar chart of prop_complete
# Add n_crossings as text labels
ggplot(cross1, aes(x = age_sex_class, y = prop_complete)) +
  geom_col() +
  geom_text(aes(label = round(prop_complete, 3)), vjust = -0.6, size = 3.5) +
  scale_y_continuous(limits = c(0, 1)) +
  labs(
    title = "Crossing completion rate by age/sex class",
    x = "Age/Sex/Class",
    y = "Proportion complete"
  ) +
  theme_bw()

```

Crossing completion rate by age/sex class

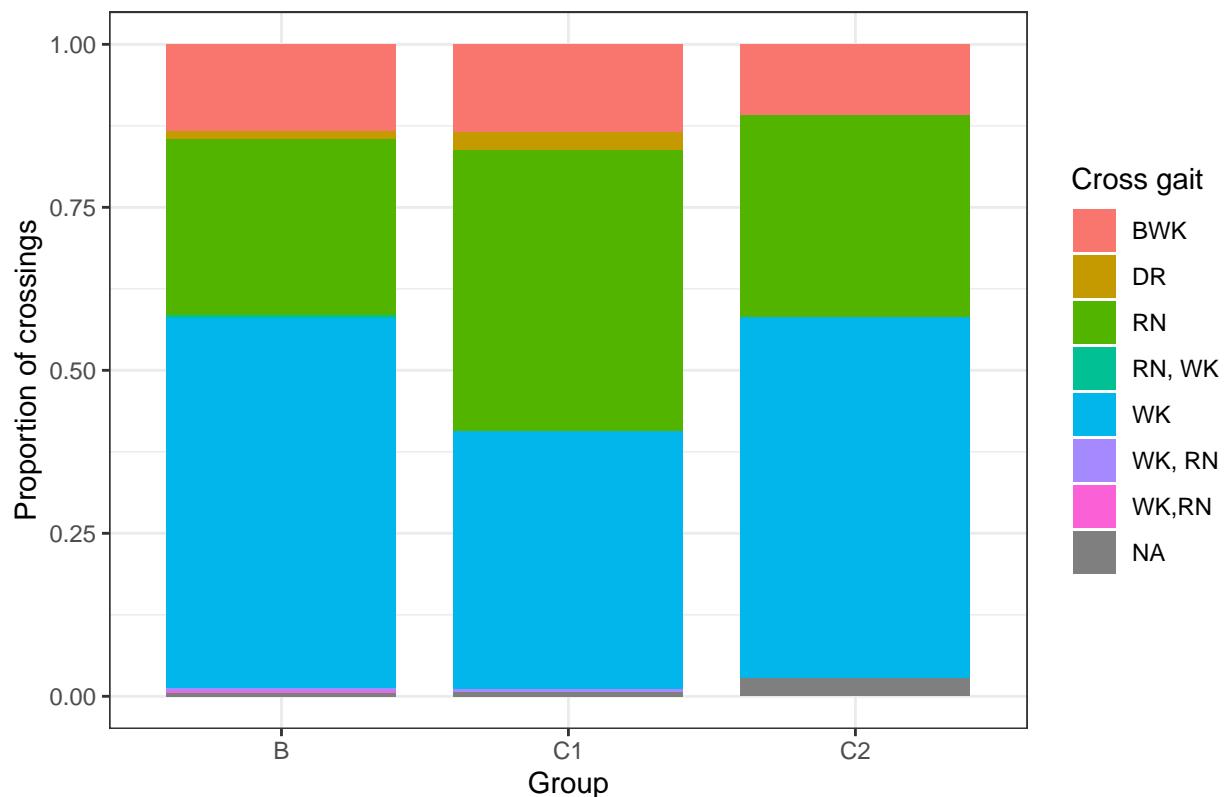


```
# add: %>% filter(!is.na(age_sex_class)) # to wrangling

# ok, let's look at how
# Gait distribution by group
cross2 <- crossings %>%
  group_by(group, cross_gait) %>%
  summarise(n = n(), .groups = "drop") %>%
  group_by(group) %>%
  mutate(prop = n / sum(n))

ggplot(cross2, aes(x = group, y = prop, fill = cross_gait)) +
  geom_col(width = 0.8) +
  scale_y_continuous(limits = c(0, 1)) +
  labs(
    title = "Crossing gait distribution by group",
    x = "Group",
    y = "Proportion of crossings",
    fill = "Cross gait"
  ) +
  theme_bw()
```

Crossing gait distribution by group



5. Bonus content

```
# heat map;
ggplot(penguins, aes(x = bill_length_mm, y = bill_depth_mm)) +
  geom_density2d_filled(color = "black", linewidth = 0.2) +
  facet_wrap(~ spp) +
  labs(
    title = "Overplotting solution: Heat Mat",
    subtitle = "Colors depict the number of points that fall in that region",
    x = "Bill length (mm)",
    y = "Bill depth (mm)",
    fill = "Count"
  ) +
  theme_bw()
```

Overplotting solution: Heat Mat

Colors depict the number of points that fall in that region

