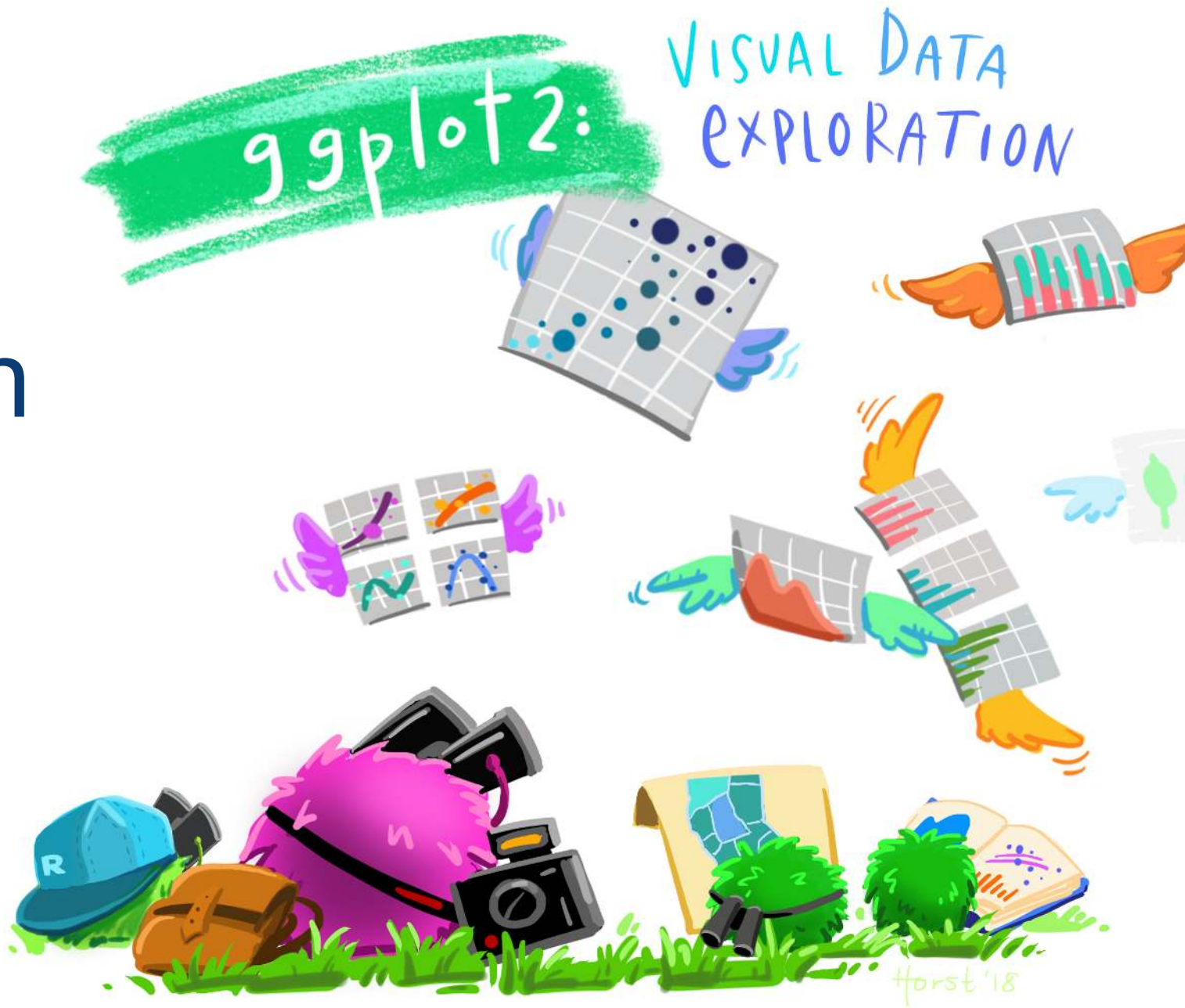


NRI 7350

Data Exploration

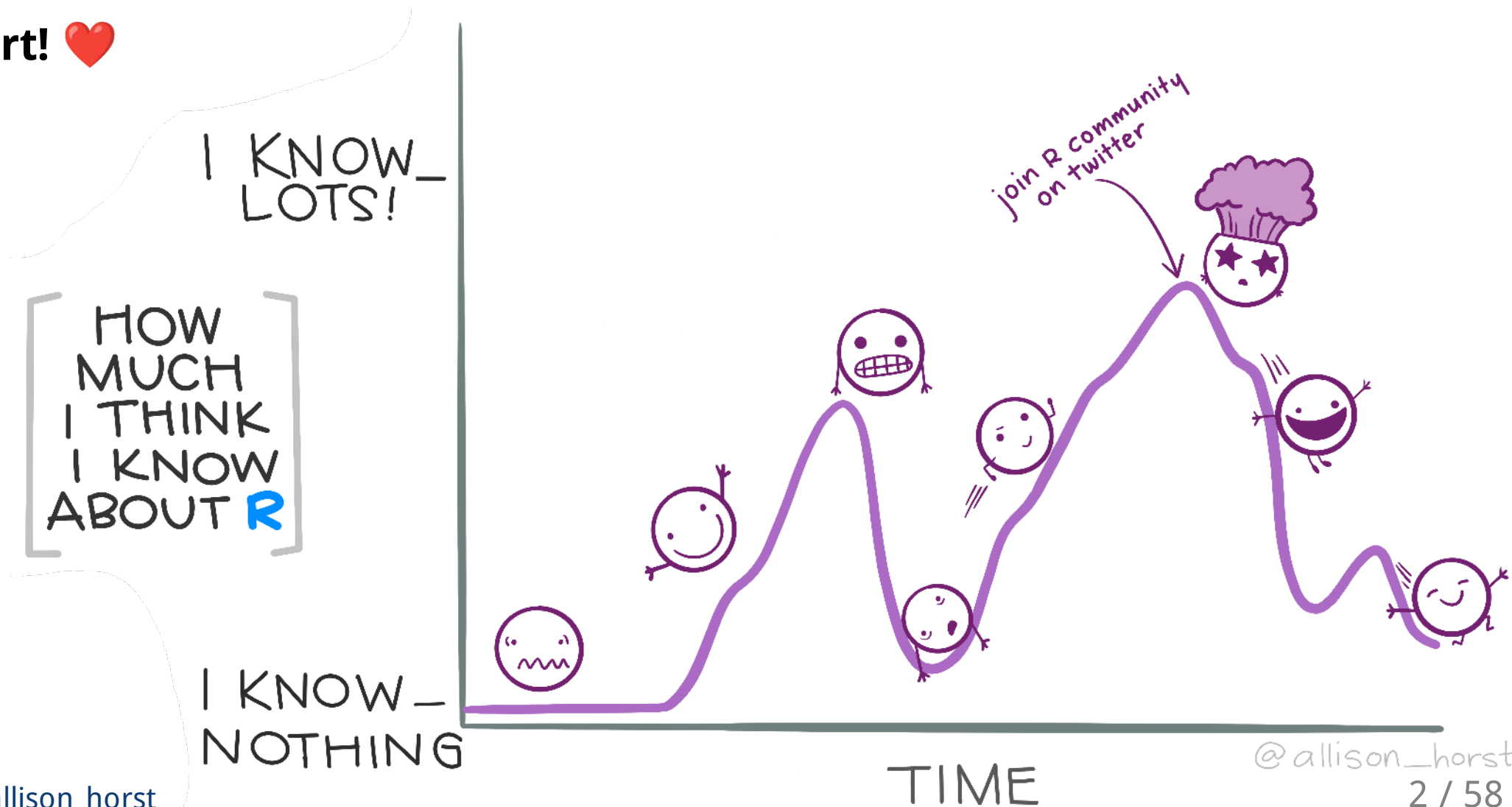
Also **GGally**, **skimr**, **dplyr**, and **moments**



Artwork by [@allison_horst](https://twitter.com/allison_horst)

How Are we Doing?

Take Heart! ❤️



Learning to Compile Reports!

Open RStudio

Open your NRI project

Open a **old** script (if it's not already open)

'Files' pane (lower right) > Click on the script name

lab1.R x lab2.R x

Source on Save

1 `library(tidyverse)`
2 `library(palmerpenguins)`
3
4 `ggplot(data = penguins, aes(x = island, y = bill_depth_mm)) +`
5 `geom_boxplot(colour = "blue")`
6
7 `ggplot(data = penguins, aes(x = sex, y = flipper_length_mm, fill = sex)) +`
8 `geom_boxplot() +`
9 `facet_wrap(~ species)`
10
11 `penguins`
12

Compile Report from R Script

Create a standalone report that contains the code and output from your R script.

For more information on compiling reports, see the documentation at [Compiling Reports from R Scripts](#)

Report output format:

HTML

Compile

Cancel

Environment History Connections Tutorial

Import Dataset 176 MiB

List

R Global Environment

Environment is empty

12:1 (Top Level)

Console Terminal Jobs

R 4.1.1 ~/Projects/Teaching/NRI_Stats_lab/

R is free software and comes with ABSOLUTELY NO WARRANTY.
You are welcome to redistribute it under certain conditions.
Type 'license()' or 'licence()' for distribution details.

Natural language support but running in an English locale

R is a collaborative project with many contributors.
Type 'contributors()' for more information and
'citation()' on how to cite R or R packages in publications.

Type 'demo()' for some demos, 'help()' for on-line help, or
'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.

> |

Files Plots Packages Help Viewer

Zoom Export


```
lab1.R x lab2.R x
Source on Save Run Source
1 library(tidyverse)
2 library(palmerpenguins)
3
4 ggplot(data = penguins, aes(x = island, y = bill_depth_mm)) +
5   geom_boxplot(colour = "blue")
6
7 ggplot(data = penguins, aes(x = sex, y = flipper_length_mm, fill = sex)) +
8   geom_boxplot() +
9   facet_wrap(~ species)
10
11 penguins
12
```

12:1 (Top Level) R Script

Console Terminal R Markdown Jobs

.../NRI_Stats_lab/lab2.R

|.....| 100%
ordinary text without R code

output file: lab2.knit.md

```
/usr/lib/rstudio/bin/pandoc/pandoc +RTS -K512m -RTS lab2.knit.md --to html4 --from markdown+autolink_bare_uris+tex_math_single_backslash --output lab2.html --lua-filter /home/steffi/R/x86_64-pc-linux-gnu-library/4.1/rmarkdown/rmarkdown/lua/pagebreak.lua --lua-filter /home/steffi/R/x86_64-pc-linux-gnu-library/4.1/rmarkdown/rmarkdown/lua/latex-div.lua --self-contained --variable bs3=TRUE --standalone --section-divs --template /home/steffi/R/x86_64-pc-linux-gnu-library/4.1/rmarkdown/rmd/h/default.html --no-highlight --variable highlightjs=1 --variable theme=bootstrap --include-in-header /tmp/Rtmp62Bdb7/rmarkdown-str69e678ba4c54.html --mathjax --variable 'mathjax-url:http://mathjax.rstudio.com/latest/MathJax.js?config=TeX-AMS-MML_HTMLorMML'
```

Output created: lab2.html

Environment History Connections Tutorial

Import Dataset 177 MiB

R Global Environment

Environment is empty

Files Plots Packages Help Viewer

New Folder Delete Rename More

Home > Projects > Teaching > NRI_Stats_lab

Name	Size	Modified
..		
.Rproj.user		
lab1.R	387 B	Sep 15, 2021, 12:39
lab2.html	808.3 KB	Sep 27, 2021, 11:33
lab2.R	267 B	Sep 27, 2021, 11:33
NRI_Stats_lab.Rproj	205 B	Sep 27, 2021, 11:32

5 / 58

Compile script into report

This is **all I need**:

- Your assignment answers
- This html report

lab2.R

steffi

2021-09-27

```
library(tidyverse)
```

```
## — Attaching packages — tidyverse 1.3.1 —
```

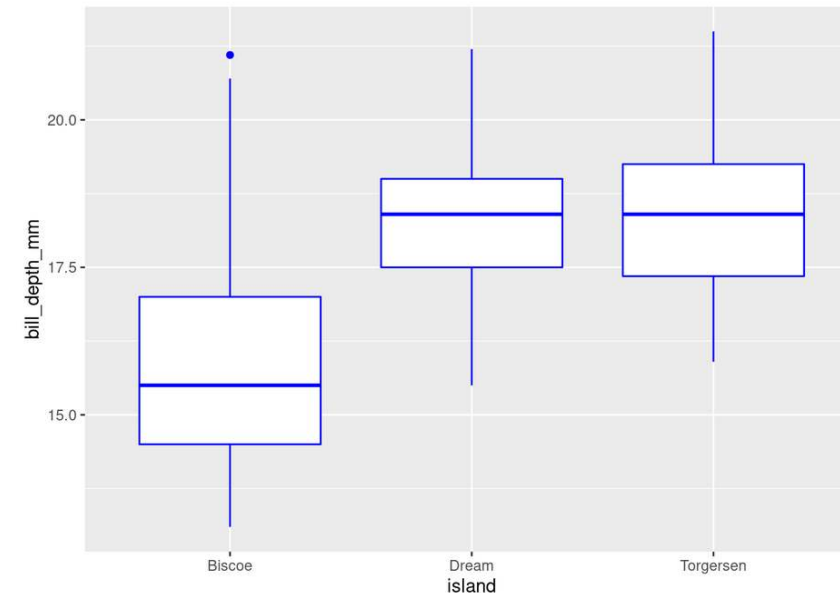
```
## ✓ ggplot2 3.3.5   ✓ purrr   0.3.4  
## ✓ tibble  3.1.4   ✓ dplyr  1.0.7  
## ✓ tidyr   1.1.3   ✓ stringr 1.4.0  
## ✓ readr   2.0.0   ✓ forcats 0.5.1
```

```
## — Conflicts — tidyverse_conflicts() —  
## x dplyr::filter() masks stats::filter()  
## x dplyr::lag()    masks stats::lag()
```

```
library(palmerpenguins)
```

```
# My first plot  
ggplot(data = penguins, aes(x = island, y = bill_depth_mm)) +  
  geom_boxplot(colour = "blue")
```

```
## Warning: Removed 2 rows containing non-finite values (stat_boxplot).
```



However!

If you want to take it further:

- You can use Markdown formatting to make things look nice
- You just need to put `#'` in front of any Markdown notation or text (DEMO)

Markdown

- **`**bold**`**
- *`*italic*`*
- `#`, `##`, `###` for headings (first-, second-, third-level)

Familiar with Rmd?

Think of this as the mirror image:

Instead of marking **what is** R code, you mark **what is NOT** R code

```
lab1.R x lab2_formatted.R x lab2.R x
← → Source on Save Run Source
1 #' ---
2 #' title: Lab 2
3 #' output:
4 #'   html_document:
5 #'     toc: true
6 #' ---
7 #'
8 #' # Background
9 #' This script is all about the plots I made in lab.
10 #' I'm leaving my self some notes here so I don't forget everything I learned!
11 #'
12 #' # Load packages
13 library(tidyverse)
14 library(palmerpenguins)
15
16 #' # Plots
17 #'
18 #' ## First plot
19 ggplot(data = penguins, aes(x = island, y = bill_depth_mm)) +
20   geom_boxplot(colour = "blue")
21
22 #' ## Second plot
23 ggplot(data = penguins, aes(x = sex, y = flipper_length_mm, fill = sex)) +
24   geom_boxplot() +
25   facet_wrap(~ species)
```


Further reading

- [R Markdown Cookbook](#)
 - Chapter 3.3 [Render an R script to a report](#)
- [R for Data Science](#)
 - Chapter 27 [Rmarkdown](#)

Lab 2

steffi

2021-09-27

- [Background](#)
- [Load packages](#)
- [Plots](#)
 - [First plot](#)
 - [Second plot](#)
- [Information for reproducibility](#)

Background

This script is all about the plots I made in lab. I'm leaving my self some notes here so I don't forget everything I learned!

Load packages

```
library(tidyverse)
```

```
## — Attaching packages — tidyverse 1.3.1 —
```

```
## ✓ ggplot2 3.3.5    ✓ purrr  0.3.4
## ✓ tibble  3.1.4    ✓ dplyr  1.0.7
## ✓ tidyr   1.1.3    ✓ stringr 1.4.0
## ✓ readr   2.0.0    ✓ forcats 0.5.1
```

```
## — Conflicts — tidyverse_conflicts() —
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()
```

```
library(palmerpenguins)
```

Getting started (again)

Hoping you can work with some of your own data today!

Open RStudio

Open your NRI project

Open your data-loading script:

'Files' Pane > Click on script name

RUN YOUR SCRIPT

Make sure to load packages at the top:

`library(tidyverse)`

`library(palmerpenguins)`

(if working with penguins today)

Exploring everything at once

Visualize with `ggpairs()`

- From **GGally** package
 - **Caution!** If you have a lot of columns, `select()` only a few to work with

```
library(GGally)

penguins_sub <- select(penguins, -sex, -island, -year)
ggpairs(penguins_sub)
```

Side Note: tidyverse functions

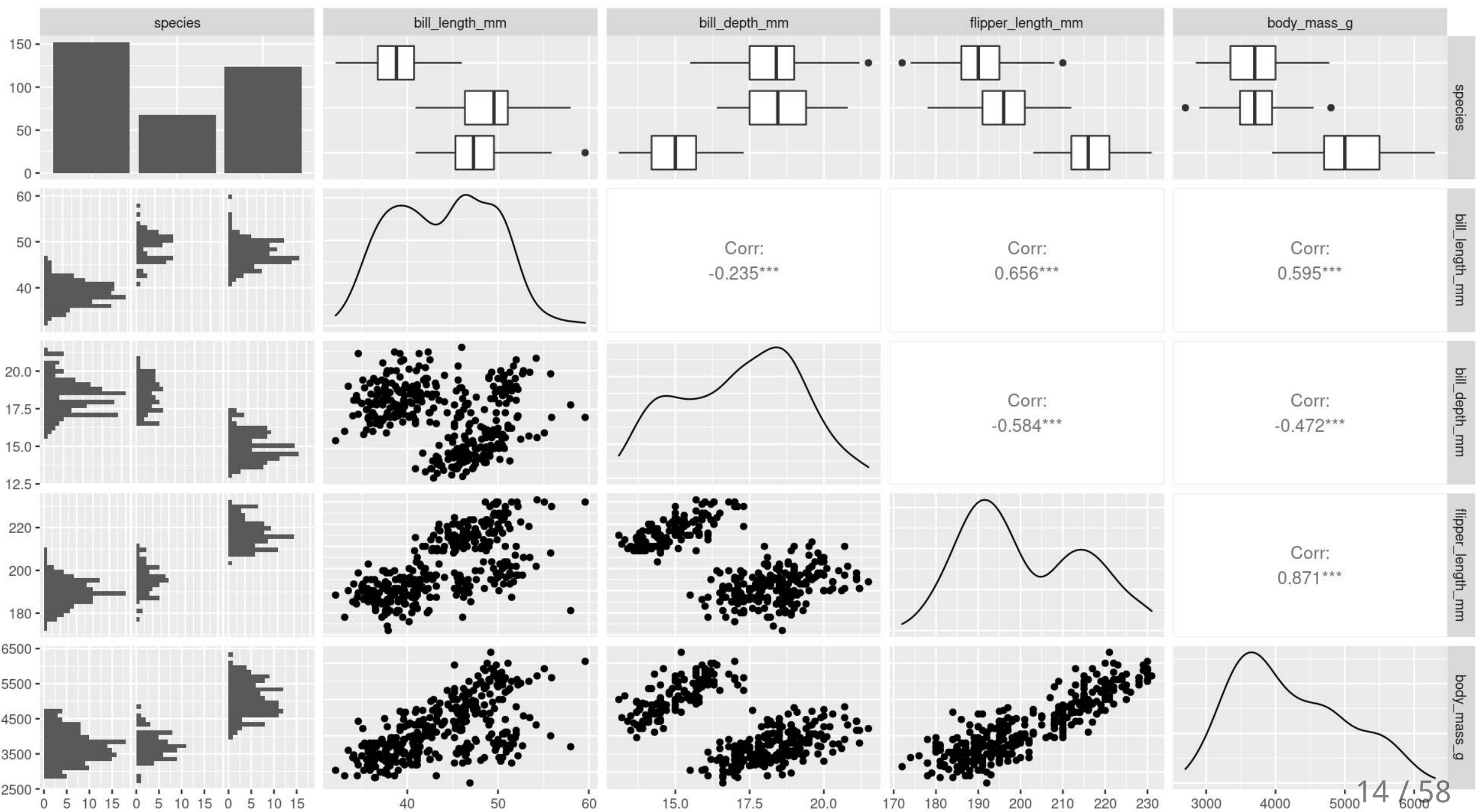
- From **GGally**
 - **Caution!** If you have a lot of columns, **select()** only a few to work with

```
library(GGally)

penguins_sub <- select(penguins, -sex, -island, -year)
ggpairs(penguins_sub)
```

select()

- **tidyverse** functions always start with the **data**, followed by other arguments
- you can reference any **column** from '**data**'
- **select()** chooses columns to keep or to remove (with **-**)

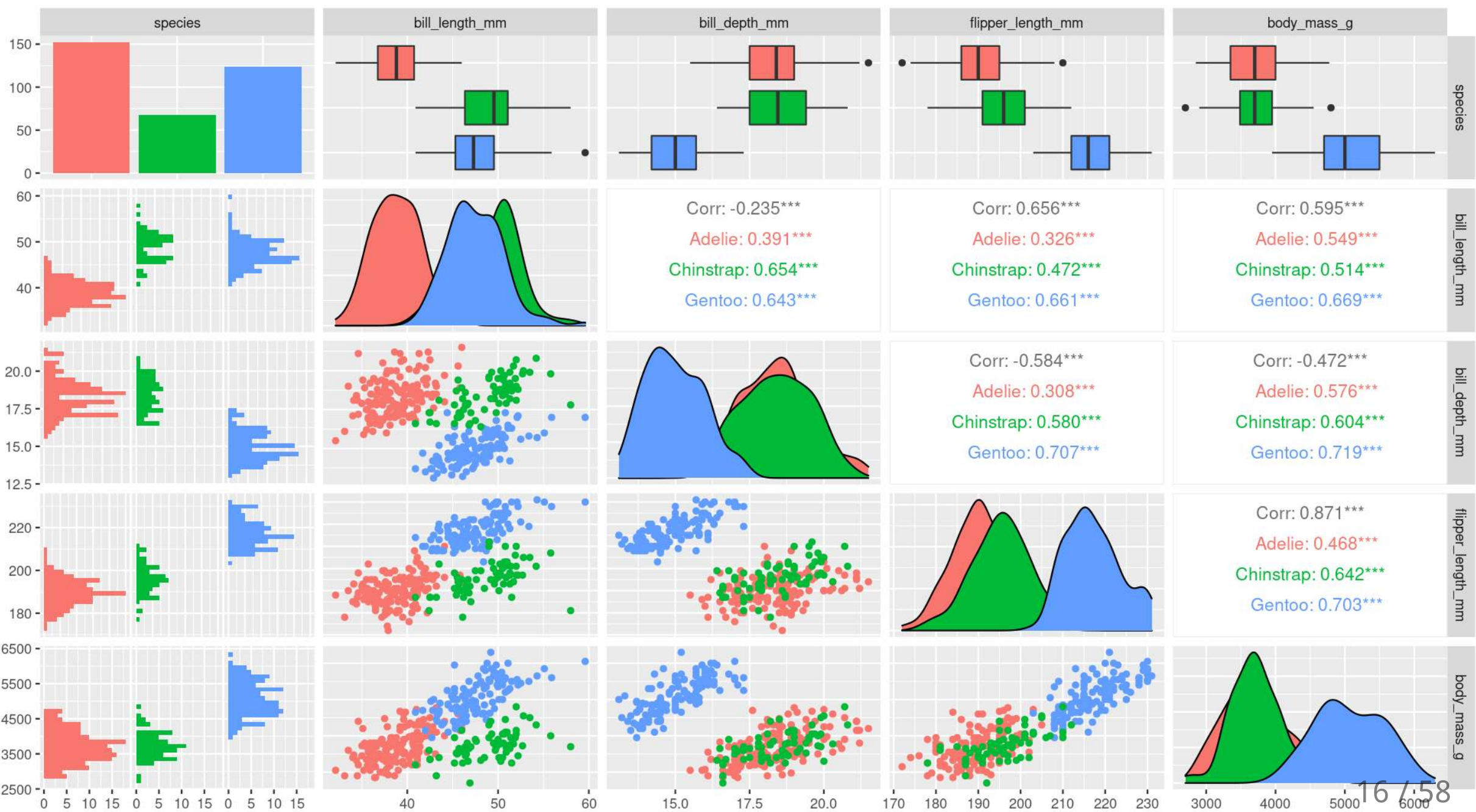


Visualize with `ggpairs()`

```
library(GGally)

ggpairs(select(penguins, -sex, -island, -year), aes(colour = species))
```

`ggpairs()` builds on **`ggplot()`** so we can use an **`aes()`** specification



Visualize with `ggpairs()`

```
library(GGally)
penguins_sub <- select(penguins, -sex, -island, -year)

ggpairs(penguins_sub)

ggpairs(penguins_sub, aes(colour = species))
```

Your Turn!

Summarize with `skim()`

`skim()` from `skimr` package






```
library(skimr)
skim(penguins)
```

```
## — Data Summary —————
##                               Values
## Name                         penguins
## Number of rows                344
## Number of columns             8
## -----
## Column type frequency:
##   factor                       3
##   numeric                      5
## -----
## Group variables                None
##
## — Variable type: factor —————
##   skim_variable n_missing complete_rate ordered n_unique top_counts
## 1 species         0           1      FALSE         3 Ade: 152, Gen: 124, Chi: 68
## 2 island          0           1      FALSE         3 Bis: 168, Dre: 124, Tor: 52
## 3 sex             11          0.968 FALSE         2 mal: 168, fem: 165
##
```

Summarize with `skim()`

`skim()` from `skimr` package

```
library(skimr)
skim(penguins)
```

```
##
## — Variable type: factor —————
##   skim_variable n_missing complete_rate ordered n_unique top_counts
## 1 species          0           1      FALSE          3 Ade: 152, Gen: 124, Chi: 68
## 2 island           0           1      FALSE          3 Bis: 168, Dre: 124, Tor: 52
## 3 sex              11          0.968 FALSE          2 mal: 168, fem: 165
##
## — Variable type: numeric —————
##   skim_variable      n_missing complete_rate   mean      sd      p0      p25      p50      p75      p100 hist
## 1 bill_length_mm        2          0.994   43.9    5.46   32.1   39.2   44.4   48.5   59.6 
## 2 bill_depth_mm         2          0.994   17.2    1.97   13.1   15.6   17.3   18.7   21.5 
## 3 flipper_length_mm      2          0.994  201.    14.1   172    190    197    213    231 
## 4 body_mass_g           2          0.994 4202.   802.   2700   3550   4050   4750   6300 
## 5 year                  0           1    2008.    0.818 2007   2007   2008   2009   2009 
```

Summarize with `skim()`

`skim()` from `skimr` package

Your Turn!

```
library(skimr)
skim(penguins)
```

```
##
## — Variable type: factor —————
##   skim_variable n_missing complete_rate ordered n_unique top_counts
## 1 species          0           1      FALSE          3 Ade: 152, Gen: 124, Chi: 68
## 2 island           0           1      FALSE          3 Bis: 168, Dre: 124, Tor: 52
## 3 sex              11          0.968 FALSE          2 mal: 168, fem: 165
##
## — Variable type: numeric —————
##   skim_variable      n_missing complete_rate      mean      sd      p0      p25      p50      p75      p100 hist
## 1 bill_length_mm         2          0.994    43.9    5.46    32.1    39.2    44.4    48.5    59.6  ████
## 2 bill_depth_mm         2          0.994    17.2    1.97    13.1    15.6    17.3    18.7    21.5  █████
## 3 flipper_length_mm      2          0.994   201.    14.1    172     190    197    213    231  ████
## 4 body_mass_g           2          0.994  4202.   802.    2700   3550   4050   4750   6300  █████
## 5 year                  0           1    2008.    0.818  2007   2007   2008   2009   2009  ████
```


Summarize with `skim()`

`group_by` from `dplyr` package

```
penguins_sp <- group_by(penguins, species)
skim(penguins_sp)
```

Side Note: **tidyverse** functions

group_by from **dplyr** package

```
penguins_sp <- group_by(penguins, species)
skim(penguins_sp)
```

group_by()

- **tidyverse** functions always start with the **data**, followed by other arguments
- you can reference any **column** from '**data**'
- **group_by()** assigns grouping to a data frame. Here, we group **penguins** by species

Extra:

In the console look at **penguins** (type in **penguins** and hit enter), and then look at **penguins_sp** (type in **penguins_sp** and hit enter).

How does the output differ? (Hint very little! But there is one difference...)

Summarize with `skim()`

`group_by` from `dplyr` package

```
penguins_sp <- group_by(penguins, species)
skim(penguins_sp)
```

```
## — Data Summary —————
##                               Values
## Name                         skimp
## Number of rows               344
## Number of columns            8
## -----
## Column type frequency:
##   factor                      2
##   numeric                     5
## -----
## Group variables               species
##
## — Variable type: factor —————
```

	skim_variable	species	n_missing	complete_rate	ordered	n_unique	top_counts
## 1	island	Adelie	0	1	FALSE	3	Dre: 56, Tor: 52, Bis: 44
## 2	island	Chinstrap	0	1	FALSE	1	Dre: 68, Bis: 0, Tor: 0
## 3	island	Gentoo	0	1	FALSE	1	Bis: 124, Dre: 0, Tor: 0
## 4	sex	Adelie	6	0.961	FALSE	2	fem: 73, mal: 73

Summarize with `skim()`

`group_by` from `dplyr` package

```
penguins_sp <- group_by(penguins, species)
skim(penguins_sp)
```

```
##
## — Variable type: factor —————
##   skim_variable species    n_missing complete_rate ordered n_unique top_counts
## 1 island         Adelie         0           1      FALSE         3 Dre: 56, Tor: 52, Bis: 44
## 2 island         Chinstrap      0           1      FALSE         1 Dre: 68, Bis: 0, Tor: 0
## 3 island         Gentoo         0           1      FALSE         1 Bis: 124, Dre: 0, Tor: 0
## 4 sex            Adelie         6           0.961 FALSE         2 fem: 73, mal: 73
## 5 sex            Chinstrap      0           1      FALSE         2 fem: 34, mal: 34
## 6 sex            Gentoo         5           0.960 FALSE         2 mal: 61, fem: 58
##
## — Variable type: numeric —————
##   skim_variable species    n_missing complete_rate mean    sd    p0    p25    p50    p75
## 1 bill_length_mm Adelie         1           0.993   38.8   2.66  32.1  36.8  38.8  40.8
## 2 bill_length_mm Chinstrap      0           1         48.8   3.34  40.9  46.3  49.6  51.1
## 3 bill_length_mm Gentoo         1           0.992   47.5   3.08  40.9  45.3  47.3  49.6
## 4 bill_depth_mm  Adelie         1           0.993   18.3   1.22  15.5  17.5  18.4  19
## 5 bill_depth_mm  Chinstrap      0           1         18.4   1.14  16.4  17.5  18.4  19.4
## 6 bill_depth_mm  Gentoo         1           0.992   15.0   0.981 13.1  14.2  15    15.7
```

Summarize with `skim()`

`group_by` from `dplyr` package

```
penguins_sp <- group_by(penguins, species)
skim(penguins_sp)
```

Your Turn!

```
##
## — Variable type: factor —————
##   skim_variable species    n_missing complete_rate ordered n_unique top_counts
## 1 island         Adelie         0           1        FALSE         3 Dre: 56, Tor: 52, Bis: 44
## 2 island         Chinstrap      0           1        FALSE         1 Dre: 68, Bis: 0, Tor: 0
## 3 island         Gentoo         0           1        FALSE         1 Bis: 124, Dre: 0, Tor: 0
## 4 sex            Adelie         6          0.961 FALSE         2 fem: 73, mal: 73
## 5 sex            Chinstrap      0           1        FALSE         2 fem: 34, mal: 34
## 6 sex            Gentoo         5          0.960 FALSE         2 mal: 61, fem: 58
##
## — Variable type: numeric —————
##   skim_variable species    n_missing complete_rate mean    sd    p0    p25    p50    p75
## 1 bill_length_mm Adelie         1          0.993    38.8  2.66  32.1  36.8  38.8  40.8
## 2 bill_length_mm Chinstrap      0           1          48.8  3.34  40.9  46.3  49.6  51.1
## 3 bill_length_mm Gentoo         1          0.992    47.5  3.08  40.9  45.3  47.3  49.6
## 4 bill_depth_mm  Adelie         1          0.993    18.3  1.22  15.5  17.5  18.4  19
## 5 bill_depth_mm  Chinstrap      0           1          18.4  1.14  16.4  17.5  18.4  19.4
## 6 bill_depth_mm  Gentoo         1          0.992    15.0  0.981 13.1  14.2  15    15.7
```

Exploring variable by variable

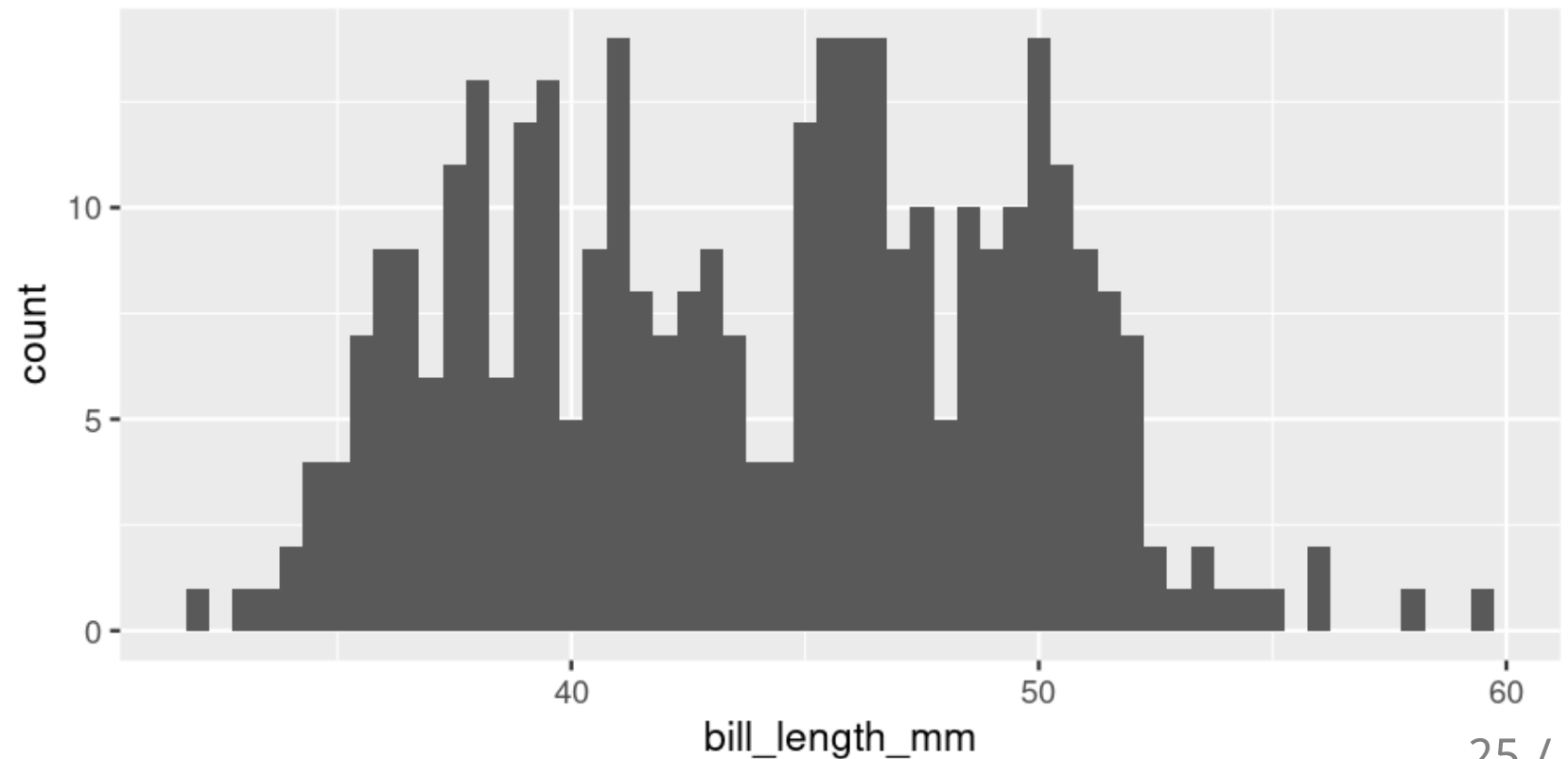
Here, use the penguins data set
(explore your own for the assignment!)

Visualize with `ggplot()`

From last week...

- Histograms

```
ggplot(data = penguins, aes(x = bill_length_mm)) +  
  geom_histogram(binwidth = 0.5)
```

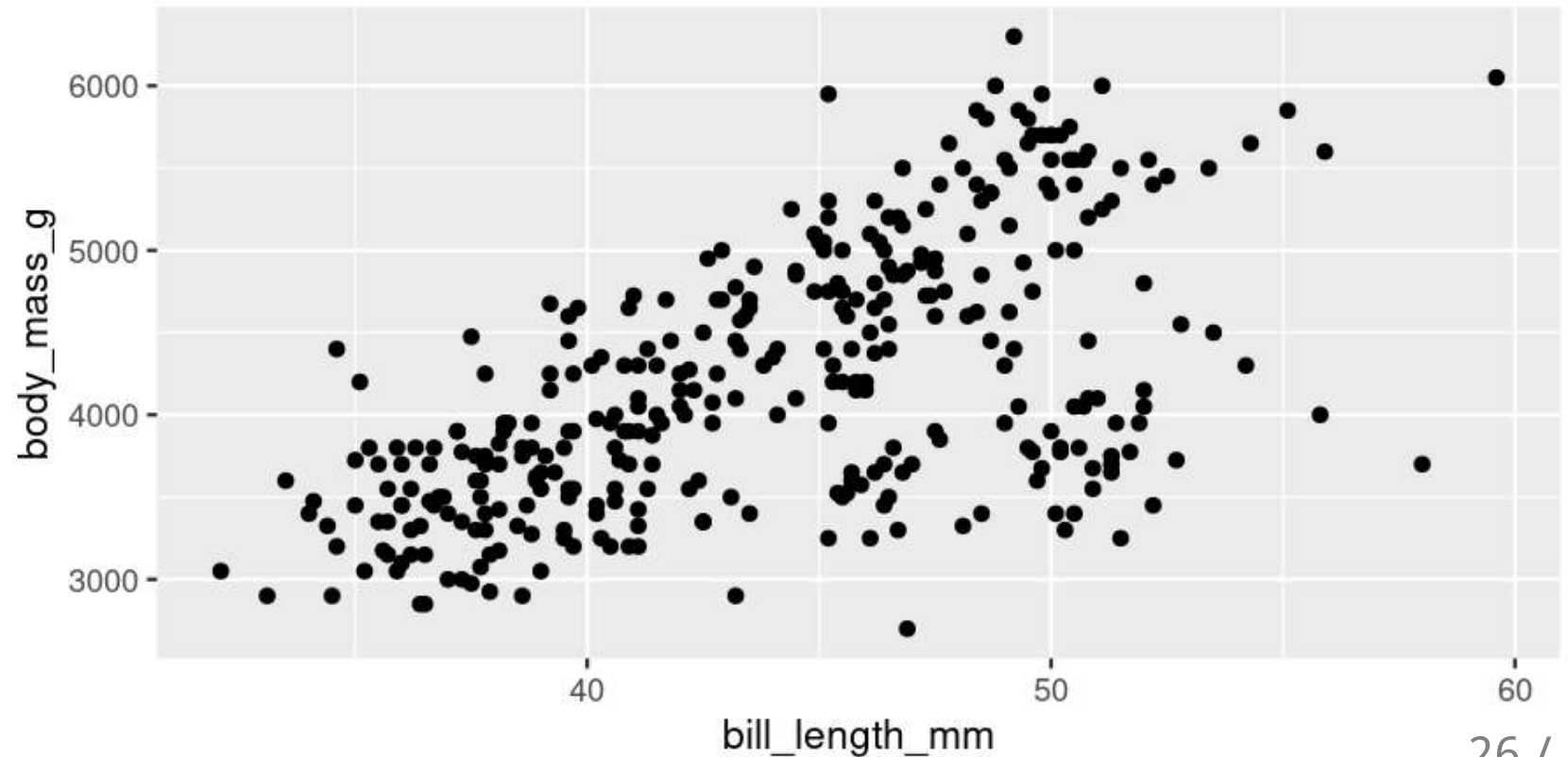


Visualize with `ggplot()`

From last week...

- Histograms
- Scatterplots

```
ggplot(data = penguins, aes(x = bill_length_mm, y = body_mass_g)) +  
  geom_point()
```

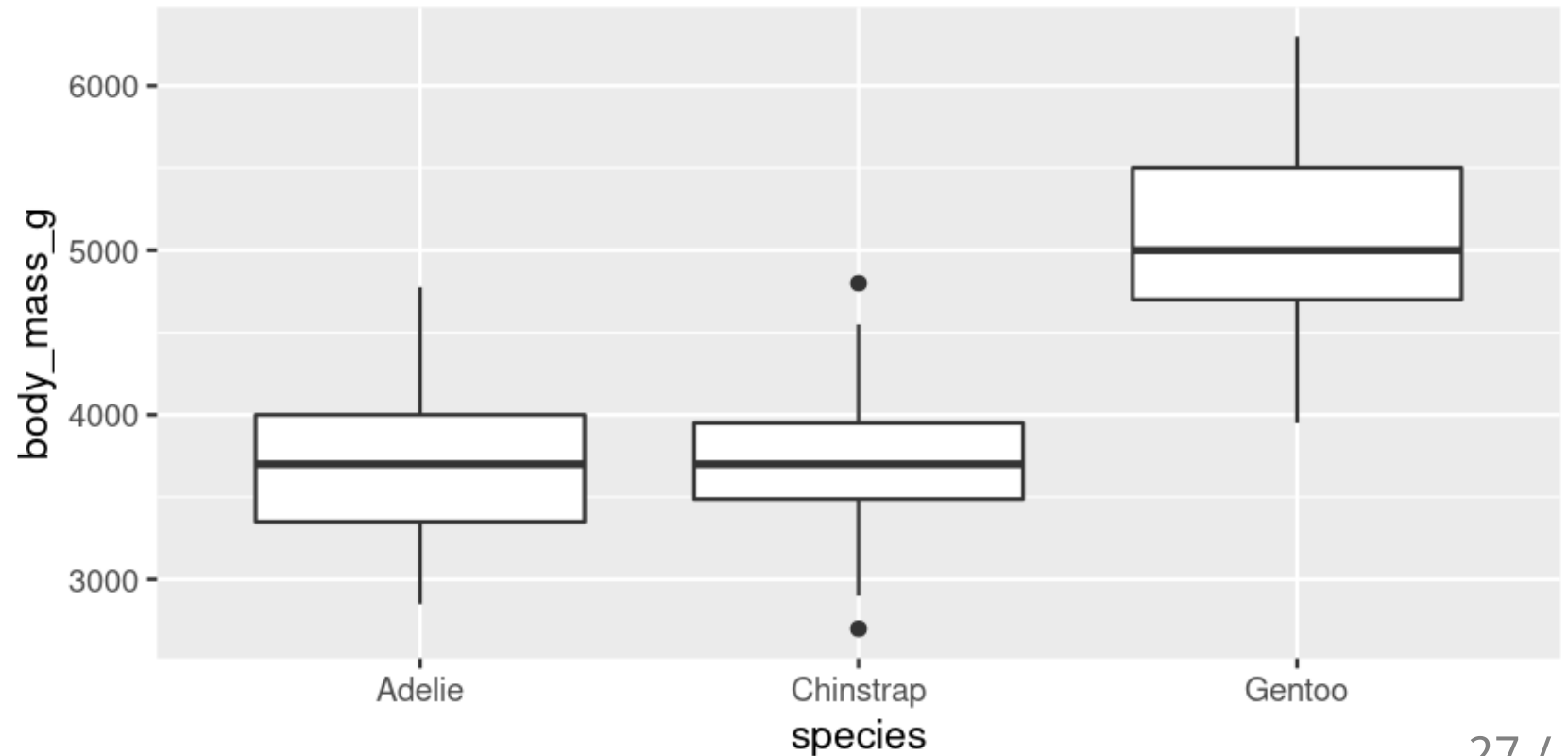


Visualize with `ggplot()`

From last week...

- Histograms
- Scatterplots
- Boxplots

```
ggplot(data = penguins, aes(x = species, y = body_mass_g)) +  
  geom_boxplot()
```

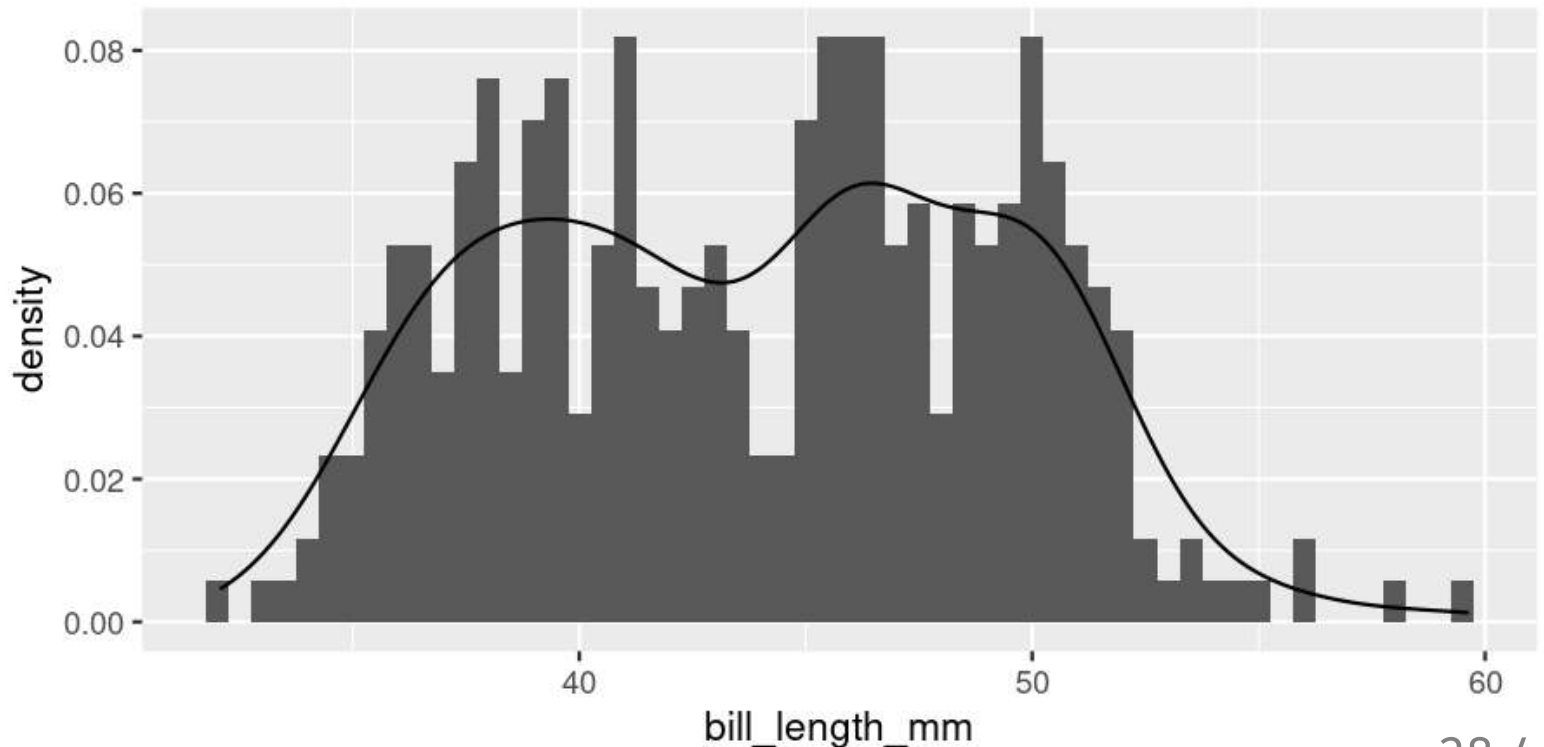


Visualize with `ggplot()`

Histogram with Density

- Default uses counts
- Here use density
`y = ..density..`
- Same as density curve
`geom_density()`
- Use to assess shape and distribution of data

```
ggplot(data = penguins, aes(x = bill_length_mm, y = ..density..)) +  
  geom_histogram(binwidth = 0.5) +  
  geom_density()
```

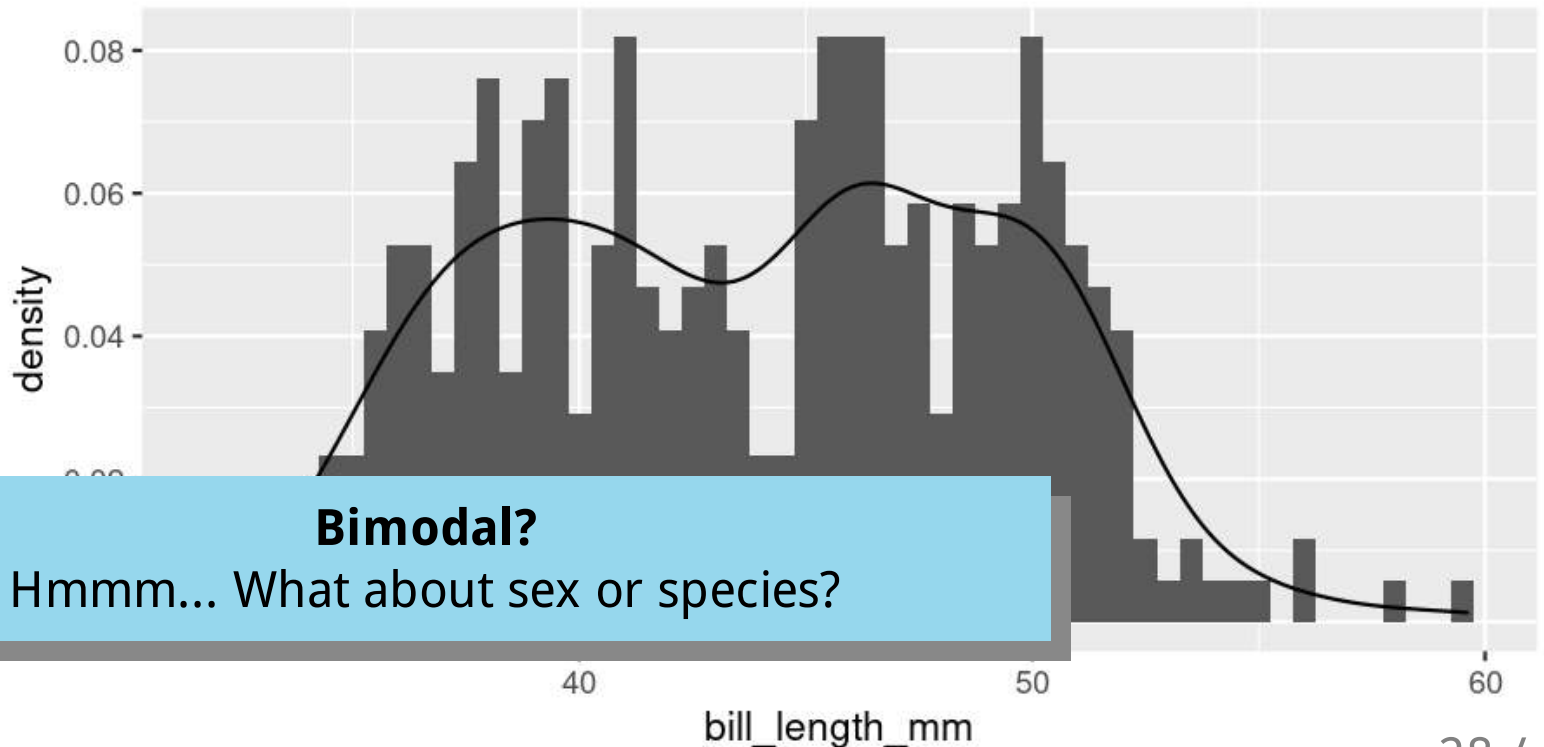


Visualize with `ggplot()`

Histogram with Density

- Default uses counts
- Here use density
`y = ..density..`
- Same as density curve
`geom_density()`
- Use to assess shape and distribution of data

```
ggplot(data = penguins, aes(x = bill_length_mm, y = ..density..)) +  
  geom_histogram(binwidth = 0.5) +  
  geom_density()
```

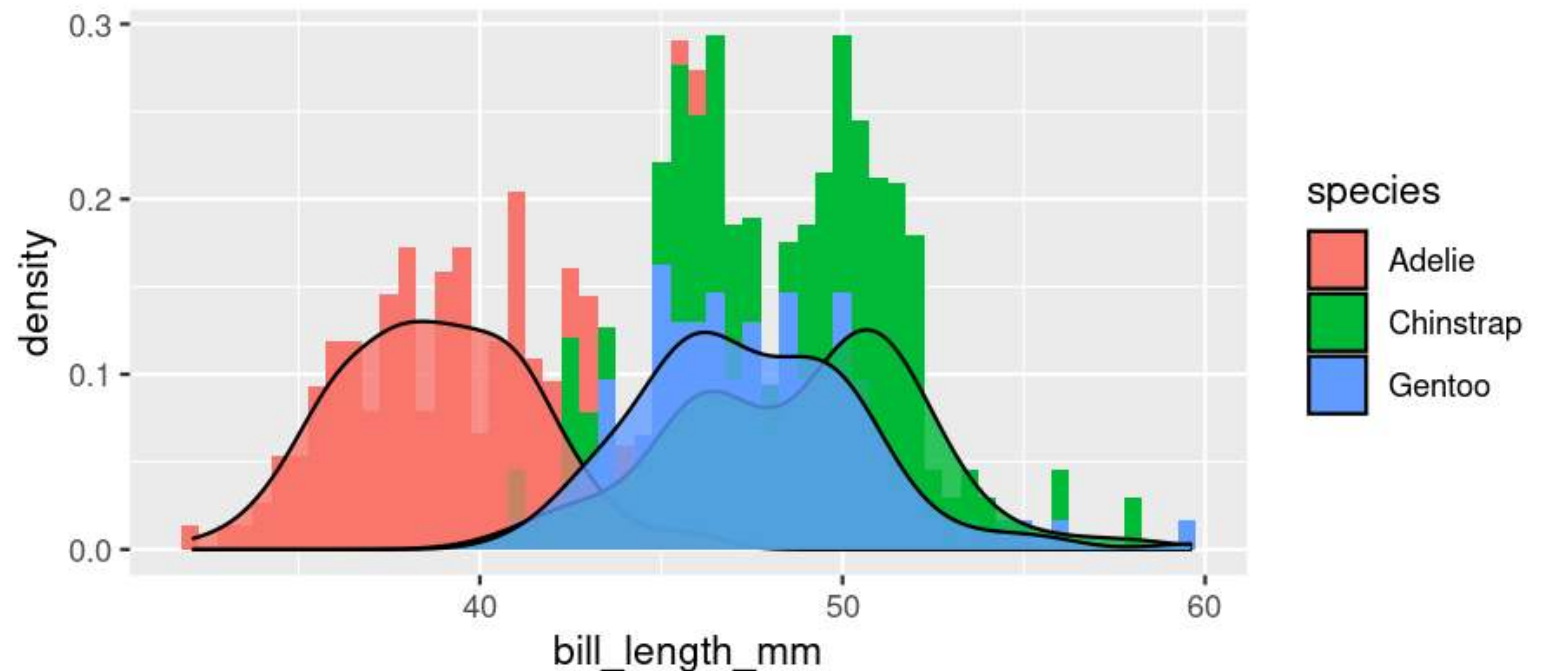


Visualize with `ggplot()`

Histogram with Density

- Default uses counts
- Here use density
`y = ..density..`
- Same as density curve
`geom_density()`
- Use to assess shape and distribution of data

```
ggplot(data = penguins, aes(x = bill_length_mm, y = ..density..,  
                             fill = species)) +  
  geom_histogram(binwidth = 0.5) +  
  geom_density(alpha = 0.8)
```

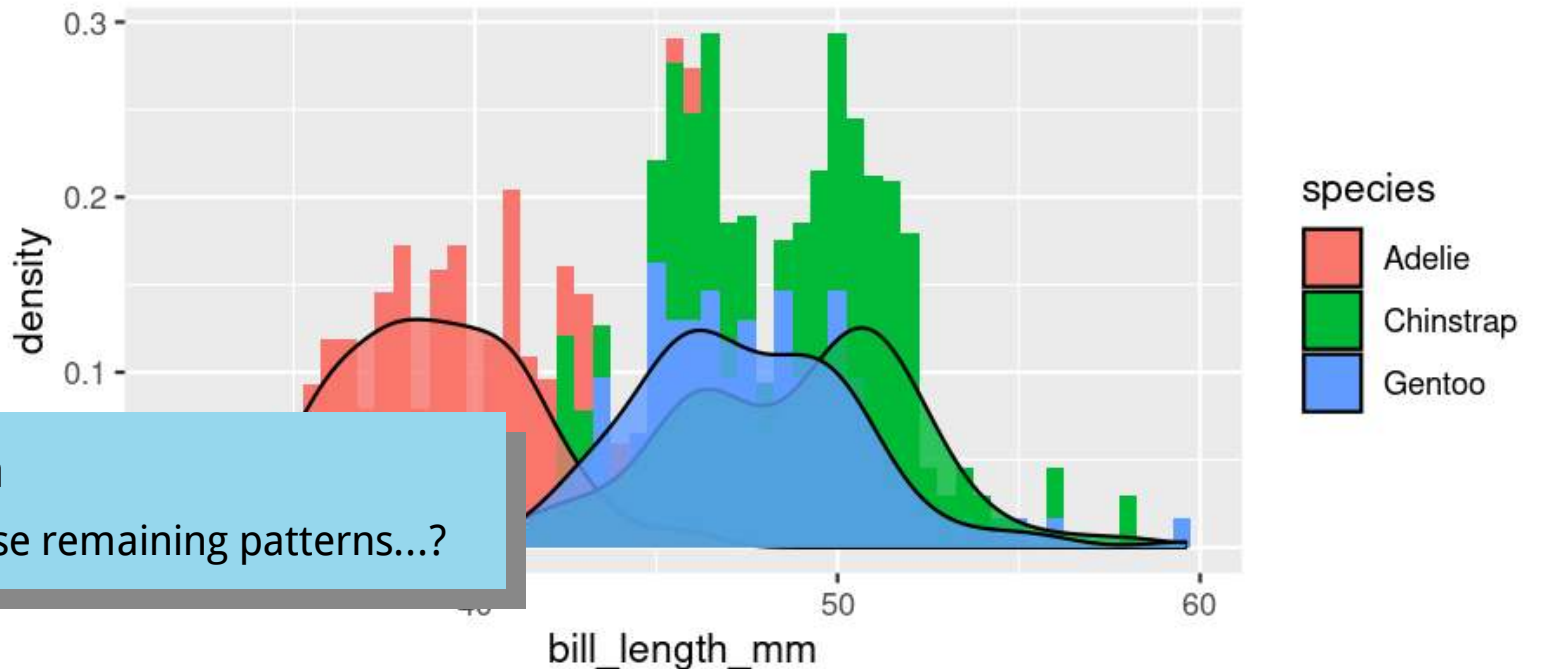


Visualize with `ggplot()`

Histogram with Density

- Default uses counts
- Here use density
`y = ..density..`
- Same as density curve
`geom_density()`
- Use to assess shape and distribution of data

```
ggplot(data = penguins, aes(x = bill_length_mm, y = ..density..,  
                             fill = species)) +  
  geom_histogram(binwidth = 0.5) +  
  geom_density(alpha = 0.8)
```



Extra

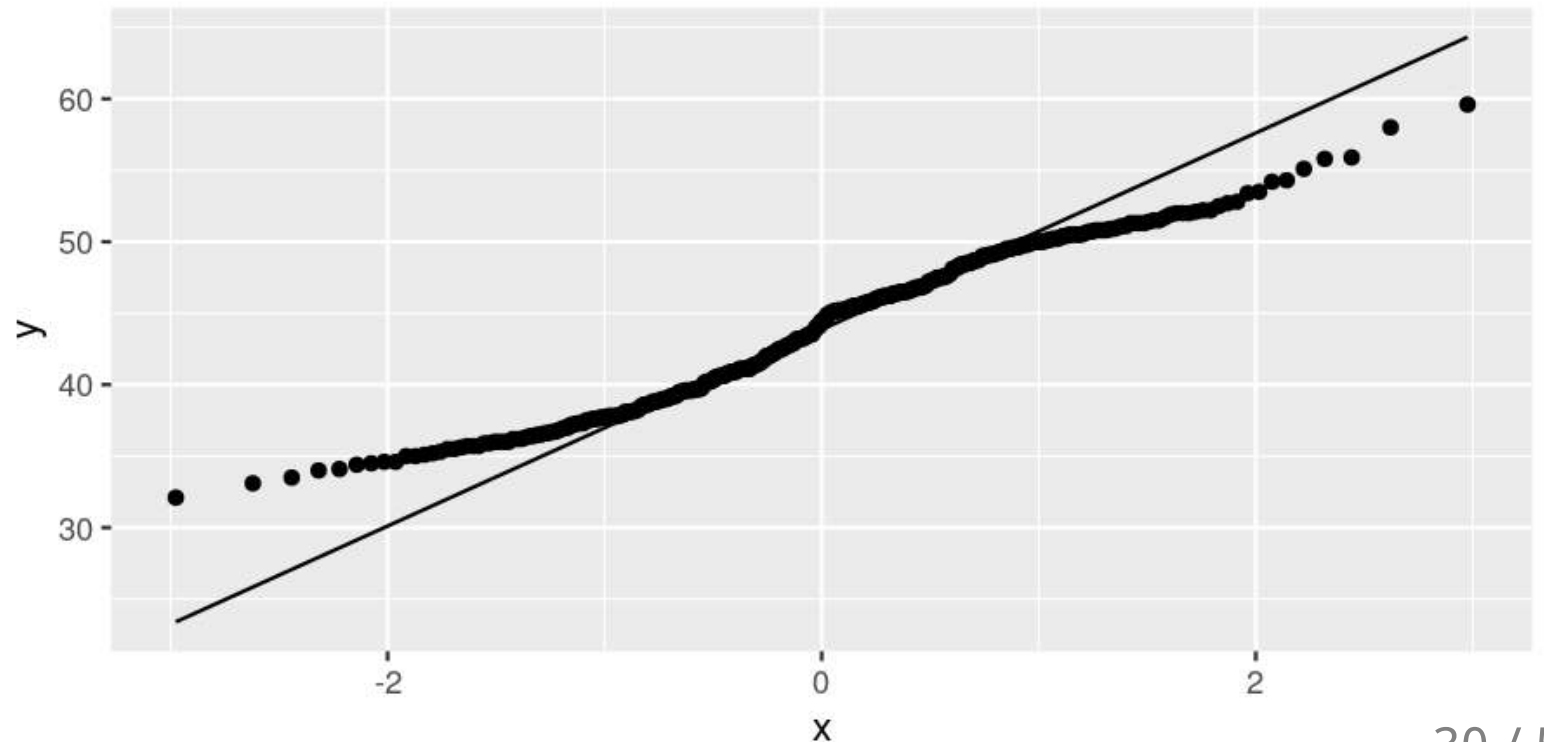
Perhaps sex plays a role in those remaining patterns...?

Visualize with `ggplot()`

QQ Norm plots

- Assess whether data follows normal distribution

```
ggplot(data = penguins, aes(sample = bill_length_mm)) +  
  stat_qq() +      # Add the points  
  stat_qq_line()   # Add the line
```

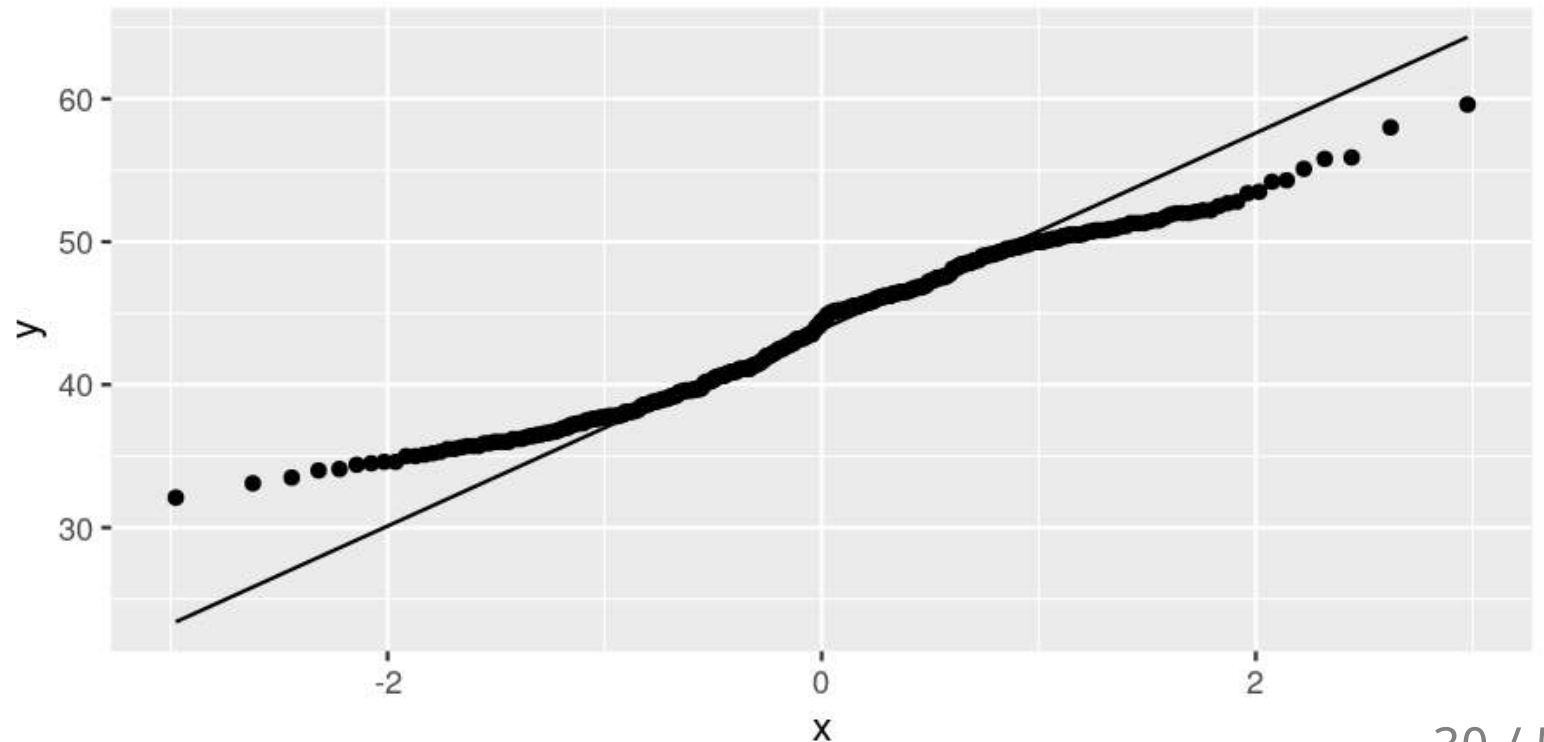


Visualize with `ggplot()`

QQ Norm plots

- Assess
follow
distribution

Here we are **NOT** assessing assumptions of normality for a model
That involves model residuals: Stay tuned for next week!



Summarize with `summarize()` Ha!

- From **dplyr** package (part of **tidyverse**)

```
summarize(penguins,  
          mean_mass = mean(body_mass_g),  
          sd_mass = sd(body_mass_g),  
          median_mass = median(body_mass_g))
```

Side Note: **tidyverse** functions

- From **dplyr** package (part of **tidyverse**)

```
summarize(penguins,  
  mean_mass = mean(body_mass_g),  
  sd_mass = sd(body_mass_g),  
  median_mass = median(body_mass_g))
```

summarize()

- **tidyverse** functions always start with the **data**, followed by **other arguments**
- you can reference any **column** from '**data**'
- **summarize()** creates a data frame with **new columns** (summarizes your data)

Summarize with `summarize()`

- From **dplyr** package (part of **tidyverse**)

```
summarize(penguins,  
          mean_mass = mean(body_mass_g),  
          sd_mass = sd(body_mass_g),  
          median_mass = median(body_mass_g))
```

```
## # A tibble: 1 × 3  
##   mean_mass sd_mass median_mass  
##   <dbl>    <dbl>      <int>  
## 1      NA      NA         NA
```

Summarize with `summarize()`

- From **dplyr** package (part of **tidyverse**)

```
summarize(penguins,  
          mean_mass = mean(body_mass_g),  
          sd_mass = sd(body_mass_g),  
          median_mass = median(body_mass_g))
```

```
## # A tibble: 1 × 3  
##   mean_mass sd_mass median_mass  
##   <dbl>    <dbl>      <int>  
## 1      NA      NA        NA
```

Why all **NAs**?

Summarize with `summarize()`

- `mean()`, `sd()`, `median()`

```
summarize(penguins,  
  mean_mass = mean(body_mass_g, na.rm = TRUE),  
  sd_mass = sd(body_mass_g, na.rm = TRUE),  
  median_mass = median(body_mass_g, na.rm = TRUE))
```

```
## # A tibble: 1 × 3  
##   mean_mass sd_mass median_mass  
##   <dbl>    <dbl>         <dbl>  
## 1    4202.     802.         4050
```

Need to tell summary statistic functions to remove missing values

`na.rm = TRUE`

Summarize with `summarize()`

- `mean()`, `sd()`, `median()`, `quantile()`, `n()`*

```
summarize(penguins,  
  mean_mass = mean(body_mass_g, na.rm = TRUE),  
  sd_mass = sd(body_mass_g, na.rm = TRUE),  
  median_mass = median(body_mass_g, na.rm = TRUE),  
  q25_mass = quantile(body_mass_g, probs = 0.25, na.rm = TRUE),  
  n = n(), # Sample size  
  n_no_missing = sum(!is.na(body_mass_g))) # Non-missing sample size
```

```
## # A tibble: 1 × 6  
##   mean_mass sd_mass median_mass q25_mass      n n_no_missing  
##   <dbl>   <dbl>       <dbl>   <dbl> <int>      <int>  
## 1    4202.    802.        4050    3550   344        342
```

* `n()` only works *inside* `summarize()/mutate()`

Your Turn: `summarize()`

Calculate summary statistics for **Bill Length**

```
summarize(penguins,  
  mean_bill_length_mm = mean(bill_length_mm),  
  sd_bill_length_mm = sd(bill_length_mm),  
  min_bill_length_mm = min(bill_length_mm),  
  max_bill_length_mm = max(bill_length_mm),  
  q1_bill_length_mm = quantile(bill_length_mm, 0.25),  
  q3_bill_length_mm = quantile(bill_length_mm, 0.75))
```

Side Note: Removing NAs

- With arguments
 - **na.rm = TRUE** (summary stats i.e. **mean()**, **sd()**)
 - **na.action = na.exclude** (models i.e., **lm()**, **lmer()**)
- You can remove all **NAs** from your data (**drop_na()**)
- You can selectively remove **NAs** from your data (**filter()**)

Side Note: Removing NAs

Remove all NAs

- This removes **every** row that has an **NA** in **any** column
- **drop_na()** function from **tidyr** package (part of **tidyverse**)

```
penguins_no_na <- drop_na(penguins)
```

- Consider removing columns with lots of **NAs** first (assuming you don't need them)

```
penguins_no_na <- select(penguins, -sex)  
penguins_no_na <- drop_na(penguins_no_na)
```

Side Side Note: **tidyverse** functions

- From **tidyr** package (part of **tidyverse**)

```
penguins_no_na <- drop_na(penguins)
```

drop_na()

- **tidyverse** functions always start with the **data**, followed by other arguments
- here, there are no other arguments

Side Note: Removing NAs

Selectively remove NAs with `filter()`

- From **dplyr** package (part of tidyverse)

```
filter(penguins, !is.na(body_mass_g))
```

- **is.na()** checks if there is an **NA** and returns **TRUE** if so
- **!** turns a **TRUE** into a **FALSE**
- **filter()** only keeps rows that are **TRUE**
- **Thus** any row with an **NA** in **body_mass_g** is removed

Side Side Note: **tidyverse** functions

- From **dplyr** package (part of **tidyverse**)

```
filter(penguins, !is.na(body_mass_g))
```

filter()

- **tidyverse** functions always start with the **data**, followed by other arguments
- you can reference any **column** from '**data**'
- **filter()** keeps only rows that return **TRUE** to the logical statements

Summarize with `summarize()` (and `group_by()`)

- Can also use `group_by()` to calculate summaries by groups

```
penguins_sp <- group_by(penguins, species)
summarize(penguins_sp,
  mean_mass = mean(body_mass_g, na.rm = TRUE),
  sd_mass = sd(body_mass_g, na.rm = TRUE),
  median_mass = median(body_mass_g, na.rm = TRUE))
```

```
## # A tibble: 3 × 4
##   species    mean_mass sd_mass median_mass
##   <fct>      <dbl>    <dbl>         <dbl>
## 1 Adelie    3701.      459.         3700
## 2 Chinstrap 3733.      384.         3700
## 3 Gentoo   5076.      504.         5000
```


Summarize with `summarize()` (and `group_by()`)

- Can also use `group_by()` to calculate summaries by groups

```
penguins_sp_sex <- group_by(penguins, species, sex)
summarize(penguins_sp_sex,
  mean_mass = mean(body_mass_g, na.rm = TRUE),
  sd_mass = sd(body_mass_g, na.rm = TRUE),
  median_mass = median(body_mass_g, na.rm = TRUE))
```

```
## # A tibble: 8 × 5
## # Groups:   species [3]
##   species    sex    mean_mass sd_mass median_mass
##   <fct>    <fct>      <dbl>   <dbl>      <dbl>
## 1 Adelie  female    3369.    269.     3400
## 2 Adelie  male     4043.    347.     4000
## 3 Adelie  <NA>     3540    477.     3475
## 4 Chinstrap female    3527.    285.     3550
## 5 Chinstrap male     3939.    362.     3950
## 6 Gentoo  female    4680.    282.     4700
## 7 Gentoo  male     5485.    313.     5500
## 8 Gentoo  <NA>     4588.    338.     4688.
```

Summarize with `summarize()` (and `group_by()`)

- Can also use `group_by()` to calculate summaries by groups

```
penguins_sp_sex <- group_by(penguins, species, sex)
summarize(penguins_sp_sex,
  mean_mass = mean(body_mass_g, na.rm = TRUE),
  sd_mass = sd(body_mass_g, na.rm = TRUE),
  median_mass = median(body_mass_g, na.rm = TRUE))
```

Where are the decimal points?

```
## # A tibble: 8 × 5
## # Groups:   species
##   species sex    mean_mass sd_mass median_mass
##   <fct>   <fct>    <dbl>   <dbl>      <dbl>
## 1 Adelie  female    3369.    269.      3400
## 2 Adelie  male     4043.    347.      4000
## 3 Adelie  <NA>     3540     477.      3475
## 4 Chinstrap female    3527.    285.      3550
## 5 Chinstrap male     3939.    362.      3950
## 6 Gentoo  female    4680.    282.      4700
## 7 Gentoo  male     5485.    313.      5500
## 8 Gentoo  <NA>     4588.    338.      4688.
```

Side Note: Where are the decimal points?

- **tibble** hides them for easy viewing

```
penguins_sum <- summarize(penguins_sp_sex,  
                           mean_mass = mean(body_mass_g, na.rm = TRUE),  
                           sd_mass = sd(body_mass_g, na.rm = TRUE),  
                           median_mass = median(body_mass_g, na.rm = TRUE))  
  
penguins_sum
```

```
## # A tibble: 8 × 5  
## # Groups:   species [3]  
##   species    sex  mean_mass sd_mass median_mass  
##   <fct>    <fct>    <dbl>   <dbl>      <dbl>  
## 1 Adelie  female    3369.    269.     3400  
## 2 Adelie  male      4043.    347.     4000  
## 3 Adelie  <NA>     3540     477.     3475  
## 4 Chinstrap female    3527.    285.     3550  
## 5 Chinstrap male      3939.    362.     3950  
## 6 Gentoo  female    4680.    282.     4700  
## 7 Gentoo  male      5485.    313.     5500  
## 8 Gentoo  <NA>     4588.    338.     4688.
```

Side Note: Where are the decimal points?

- **tibble** hides them for easy viewing

```
penguins_sum <- summarize(penguins_sp_sex,  
                           mean_mass = mean(body_mass_g, na.rm = TRUE),  
                           sd_mass = sd(body_mass_g, na.rm = TRUE),  
                           median_mass = median(body_mass_g, na.rm = TRUE))  
  
penguins_sum
```

```
## # A tibble: 8 × 5  
## # Groups:   species [3]  
##   species sex    mean_mass sd_mass median_mass  
##   <fct>   <fct>      <dbl>   <dbl>      <dbl>  
## 1 Adelie female    3369.    269.    3400  
## 2 Adelie male     4043.    347.    4000  
## 3 Adelie <NA>     3540    477.    3475  
## 4 Chinstrap female  3527.    285.    3550  
## 5 Chinstrap male    3939.    362.    3950  
## 6 Gentoo female   4680.    282.    4700  
## 7 Gentoo male     5485.    313.    5500  
## 8 Gentoo <NA>     4588.    338.    4688.
```

Note

If you want to keep the output, you need to assign (**<-**) it to an object.

Here, **penguins_sum**

Side Note: Where are the decimal points?

- **`as.data.frame()`** to see the raw data

```
as.data.frame(penguins_sum)
```

##	species	sex	mean_mass	sd_mass	median_mass
## 1	Adelie	female	3368.836	269.3801	3400.0
## 2	Adelie	male	4043.493	346.8116	4000.0
## 3	Adelie	<NA>	3540.000	477.1661	3475.0
## 4	Chinstrap	female	3527.206	285.3339	3550.0
## 5	Chinstrap	male	3938.971	362.1376	3950.0
## 6	Gentoo	female	4679.741	281.5783	4700.0
## 7	Gentoo	male	5484.836	313.1586	5500.0
## 8	Gentoo	<NA>	4587.500	338.1937	4687.5

- Or click on the name in the Environment pane

Side Note: Where are all my data?

```
penguins
```

```
## # A tibble: 344 × 8
##   species island   bill_length_mm bill_depth_mm flipper_length_mm body_mass_g
##   <fct>   <fct>         <dbl>         <dbl>         <int>         <int>
## 1 Adelie  Torgersen         39.1          18.7          181          3750
## 2 Adelie  Torgersen         39.5          17.4          186          3800
## 3 Adelie  Torgersen         40.3          18           195          3250
## 4 Adelie  Torgersen         NA           NA           NA           NA
## 5 Adelie  Torgersen         36.7          19.3          193          3450
## 6 Adelie  Torgersen         39.3          20.6          190          3650
## 7 Adelie  Torgersen         38.9          17.8          181          3625
## 8 Adelie  Torgersen         39.2          19.6          195          4675
## 9 Adelie  Torgersen         34.1          18.1          193          3475
## 10 Adelie Torgersen         42           20.2          190          4250
## # ... with 334 more rows, and 2 more variables: sex <fct>, year <int>
```

... with 334 more rows, and 2 more variables: sex <fct>, year <int>

Side Note: Where are all my data?

```
print(penguins, n = Inf)
```

```
## # A tibble: 344 × 8
##   species    island bill_length_mm bill_depth_mm flipper_length_... body_mass_g
##   <fct>      <fct>         <dbl>         <dbl>         <int>         <int>
## 1 Adelie    Torgersen         39.1          18.7          181          3750
## 2 Adelie    Torgersen         39.5          17.4          186          3800
## 3 Adelie    Torgersen         40.3          18           195          3250
## 4 Adelie    Torgersen         NA           NA           NA           NA
## 5 Adelie    Torgersen         36.7          19.3          193          3450
## 6 Adelie    Torgersen         39.3          20.6          190          3650
## 7 Adelie    Torgersen         38.9          17.8          181          3625
## 8 Adelie    Torgersen         39.2          19.6          195          4675
## 9 Adelie    Torgersen         34.1          18.1          193          3475
## 10 Adelie   Torgersen         42           20.2          190          4250
## 11 Adelie    Torgersen         37.8          17.1          186          3300
## 12 Adelie    Torgersen         37.8          17.3          180          3700
## 13 Adelie    Torgersen         41.1          17.6          182          3200
## 14 Adelie    Torgersen         38.6          21.2          191          3800
## 15 Adelie    Torgersen         34.6          21.1          198          4400
```

Side Note: Where are all my data?

```
as.data.frame(penguins)
```

##	species	island	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g	sex	year
## 1	Adelie	Torgersen	39.1	18.7	181	3750	male	2007
## 2	Adelie	Torgersen	39.5	17.4	186	3800	female	2007
## 3	Adelie	Torgersen	40.3	18.0	195	3250	female	2007
## 4	Adelie	Torgersen	NA	NA	NA	NA	<NA>	2007
## 5	Adelie	Torgersen	36.7	19.3	193	3450	female	2007
## 6	Adelie	Torgersen	39.3	20.6	190	3650	male	2007
## 7	Adelie	Torgersen	38.9	17.8	181	3625	female	2007
## 8	Adelie	Torgersen	39.2	19.6	195	4675	male	2007
## 9	Adelie	Torgersen	34.1	18.1	193	3475	<NA>	2007
## 10	Adelie	Torgersen	42.0	20.2	190	4250	<NA>	2007
## 11	Adelie	Torgersen	37.8	17.1	186	3300	<NA>	2007
## 12	Adelie	Torgersen	37.8	17.3	180	3700	<NA>	2007
## 13	Adelie	Torgersen	41.1	17.6	182	3200	female	2007
## 14	Adelie	Torgersen	38.6	21.2	191	3800	male	2007
## 15	Adelie	Torgersen	34.6	21.1	198	4400	male	2007
## 16	Adelie	Torgersen	36.6	17.8	185	3700	female	2007
## 17	Adelie	Torgersen	38.7	19.0	195	3450	female	2007

Summarize with `summarize()`

`skewness()`, `kurtosis()`

- From **moments** package

```
library(moments)

summarize(penguins,
          skew_mass = skewness(body_mass_g, na.rm = TRUE),
          kurt_mass = kurtosis(body_mass_g, na.rm = TRUE))
```

```
## # A tibble: 1 × 2
##   skew_mass kurt_mass
##   <dbl>     <dbl>
## 1    0.468     2.27
```

1. Normal distribution, skew = 0, kurtosis = 3*
2. Remember that it's best to evaluate the distribution **both** visually and statistically

* **Excess kurtosis** would be 0 for a normal distribution, but this functions measures **kurtosis**

Summarize with `summarize()`

Confidence Intervals

- By hand!
- 95% Confidence interval ranges from $[\text{mean} - (1.96 \text{ SE})]$ to $[\text{mean} + (1.96 \text{ SE})]$
- You can also express this interval as: $\text{mean} \pm (1.96 * \text{SE})$
- Standard Errors (SE) can be calculated by SD / \sqrt{n}

```
summarize(penguins,  
  mean_mass = mean(body_mass_g, na.rm = TRUE),  
  sd_mass = sd(body_mass_g, na.rm = TRUE),  
  n = n(),  
  se_mass = sd_mass / sqrt(n),           # Calculate Standard Error  
  ci_mass = 1.96 * se_mass,              # CI margin of error  
  ci_low_mass = mean_mass - ci_mass,     # The lower range  
  ci_high_mass = mean_mass + ci_mass)    # The upper range
```

Summarize with `summarize()`

Confidence Intervals

- By hand!
- 95% Confidence interval ranges from $[\text{mean} - (1.96 \text{ SE})]$ to $[\text{mean} + (1.96 \text{ SE})]$
- You can also express this interval as: $\text{mean} \pm (1.96 * \text{SE})$
- Standard Errors (SE) can be calculated by SD / \sqrt{n}

```
## # A tibble: 1 × 7
##   mean_mass sd_mass      n se_mass ci_mass ci_low_mass ci_high_mass
##   <dbl>    <dbl> <int>  <dbl>  <dbl>      <dbl>      <dbl>
## 1    4202.    802.   344    43.2    84.7    4117.    4287.
```

Put it All Together

```
penguins_sp <- group_by(penguins, species)
summarize(penguins_sp,
  mean_mass = mean(body_mass_g, na.rm = TRUE),
  sd_mass = sd(body_mass_g, na.rm = TRUE),
  q25_mass = quantile(body_mass_g, probs = 0.25, na.rm = TRUE),
  median_mass = median(body_mass_g, na.rm = TRUE),
  q75_mass = quantile(body_mass_g, probs = 0.25, na.rm = TRUE),
  n = n(),
  n_no_missing = sum(!is.na(body_mass_g)),
  skew_mass = skewness(body_mass_g, na.rm = TRUE),
  kurt_mass = kurtosis(body_mass_g, na.rm = TRUE),
  se_mass = sd_mass / sqrt(n),
  ci_mass = 1.96 * se_mass,
  ci_low_mass = mean_mass - ci_mass,
  ci_high_mass = mean_mass + ci_mass)
```

Put it All Together

##	species	mean_mass	sd_mass	q25_mass	median_mass	q75_mass	n	n_no_missing	skew_mass	kurt_mass
## 1	Adelie	3700.662	458.5661	3350.0	3700	3350.0	152	151	0.28249381	2.405611
## 2	Chinstrap	3733.088	384.3351	3487.5	3700	3487.5	68	68	0.24194125	3.463681
## 3	Gentoo	5076.016	504.1162	4700.0	5000	4700.0	124	123	0.06878276	2.257871
##	se_mass	ci_mass	ci_low_mass	ci_high_mass						
## 1	37.19462	72.90146	3627.761	3773.564						
## 2	46.60747	91.35065	3641.738	3824.439						
## 3	45.27097	88.73111	4987.285	5164.747						

Put it All Together (Advanced!)

`pivot_longer()` transposes data

- from **tidyr** package (part of **tidyverse**)

```
penguins_long <- pivot_longer(penguins,
                              cols = c(bill_length_mm, bill_depth_mm, flipper_length_mm,
                              body_mass_g),
                              names_to = "measurement", values_to = "values")

penguins_long
```

```
## # A tibble: 1,376 × 6
##   species island    sex   year measurement      values
##   <fct>   <fct>   <fct> <int> <chr>         <dbl>
## 1 Adelie  Torgersen male   2007 bill_length_mm    39.1
## 2 Adelie  Torgersen male   2007 bill_depth_mm     18.7
## 3 Adelie  Torgersen male   2007 flipper_length_mm 181
## 4 Adelie  Torgersen male   2007 body_mass_g     3750
## 5 Adelie  Torgersen female 2007 bill_length_mm    39.5
## 6 Adelie  Torgersen female 2007 bill_depth_mm     17.4
## 7 Adelie  Torgersen female 2007 flipper_length_mm 186
```

Put it All Together (Advanced!)

`pivot_longer()` transposes data

- from **tidyr** package (part of **tidyverse**)

```
penguins_long <- pivot_longer(penguins,
                              cols = c(bill_length_mm, bill_depth_mm, flipper_length_mm,
                              body_mass_g),
                              names_to = "measurement", values_to = "values")

penguins_long
```

```
## # A tibble: 1,376 × 6
##   species island    sex   year measurement      values
##   <fct>   <fct>   <fct> <int> <chr>         <dbl>
## 1 Adelie  Torgersen male   2007 bill_length_mm    39.1
## 2 Adelie  Torgersen male   2007 bill_depth_mm     18.7
## 3 Adelie  Torgersen male   2007 flipper_length_mm  181
## 4 Adelie  Torgersen male   2007 body_mass_g     3750
## 5 Adelie  Torgersen female 2007 bill_length_mm    39.5
## 6 Adelie  Torgersen female 2007 bill_depth_mm     17.4
## 7 Adelie  Torgersen female 2007 flipper_length_mm  186
```

Extra

Compare **penguins** to **penguins_long**.
Can you see what the **pivot_longer()**
function is doing?

Put it All Together (Advanced!)

```
penguins_long_sp <- group_by(penguins_long, species, measurement)

summarize(penguins_long_sp,
  mean = mean(values, na.rm = TRUE),
  sd = sd(values, na.rm = TRUE),
  q25 = quantile(values, probs = 0.25, na.rm = TRUE),
  median = median(values, na.rm = TRUE),
  q75 = quantile(values, probs = 0.25, na.rm = TRUE),
  n = n(),
  n_no_missing = sum(!is.na(values)),
  skew = skewness(values, na.rm = TRUE),
  kurt = kurtosis(values, na.rm = TRUE))
```


Put it All Together (Advanced!)

```
## `summarise()` has grouped output by 'species'. You can override using the `.groups` argument.
```

```
## # A tibble: 12 × 11
```

```
## # Groups:   species [3]
```

##	species	measurement	mean	sd	q25	median	q75	n	n_no_missing	skew	kurt
##	<fct>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<int>	<int>	<dbl>	<dbl>
## 1	Adelie	bill_depth_mm	18.3	1.22	17.5	18.4	17.5	152	151	0.318	2.90
## 2	Adelie	bill_length_mm	38.8	2.66	36.8	38.8	36.8	152	151	0.160	2.81
## 3	Adelie	body_mass_g	3701.	459.	3350	3700	3350	152	151	0.282	2.41
## 4	Adelie	flipper_length_mm	190.	6.54	186	190	186	152	151	0.0865	3.28
## 5	Chinstrap	bill_depth_mm	18.4	1.14	17.5	18.4	17.5	68	68	0.00673	2.10
## 6	Chinstrap	bill_length_mm	48.8	3.34	46.3	49.6	46.3	68	68	-0.0886	2.95

All Data vs. Variable by Variable

Depends on what you need

- **ggpairs()** and **skim()**
 - Lots of data quickly summarized and examined
 - Less easily customized (but still possible!)
- **ggplot()** and **summarize()**
 - Take a bit longer to write out
 - Very customizable
 - Can easily include stats not available in **ggpairs()** and **skim()**

Wrapping up: Further reading (all **Free!**)

- RStudio > Help > Cheatsheets > Data Transformation with dplyr
- [R for Data Science](#)
 - [Data transformation](#)
 - [Exploratory Data Analysis](#)