Module 3: Reporting, Data Wrangling and Graphing (II)

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Outline

Last module we reviewed how to tidy and plot data.

In this module, we will continue our discussion on

- Graphing (ggplot2) with a real-world dataset
- Git + Github

ggplot

- ggplot is the graphing package that goes with the tidyverse in R
- Very powerful to make a wide range of graphics
- Same pattern as tidyverse", but using"+" to connect.

How to write?

Specify the data using

```
ggplot(data = diamonds)
```

Specify the x-/y-axis,

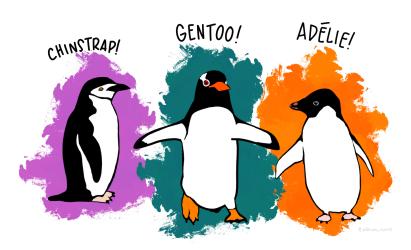
```
ggplot(data = diamonds, mapping = aes(x = cut))
```

Specify the types of plots with geom, e.g.

```
+ geom_bar()
```

Data used - palmerpenguins

The palmerpenguins is a R package with data from the Long Term Ecological Research Network. It contains two datasets for 344 penguins and 3 species of penguins from 3 islands in the Palmer Archipelago, Antarctica.



Install and load package

From what we have learned so far, how to install package? What packages are we going to use?

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```
install.packages("palmerpenguins")
```

```
library(tidyverse)
library(palmerpenguins)
```

Skim the data

How many observations? How many variables? What type?

Skim the data

How many observations? How many variables? What type?

```
head(penguins)
```

```
## # A tibble: 6 x 8
## species island bill_length_mm bill_depth_mm flipper_le
##
    <fct> <fct>
                              <dbl>
                                           <dbl>
## 1 Adelie Torgersen
                               39.1
                                            18.7
## 2 Adelie Torgersen
                             39.5
                                            17.4
                             40.3
## 3 Adelie Torgersen
                                            18
## 4 Adelie Torgersen
                               NA
                                            NΑ
## 5 Adelie Torgersen
                               36.7
                                            19.3
                                            20.6
## 6 Adelie Torgersen
                               39.3
## # i 2 more variables: sex <fct>, year <int>
```

Scatter plot

Consider a scatter plot of flipper length and body mass for species = "Adelie"

First let's prepare the data to plot (Hint: filter).

Scatter plot

Consider a scatter plot of flipper length and body mass for species = "Adelie"

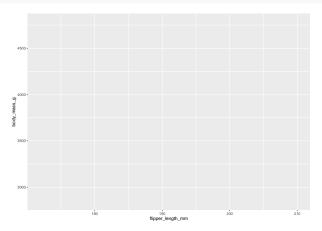
First let's prepare the data to plot (Hint: filter).

```
pdata <- penguins %>%
  filter(species == "Adelie")
```

A blank canvas

aes stands for aesthetic and tells ggplot the main characteristics of your plot (x, y, and if the color or fill vary by group)

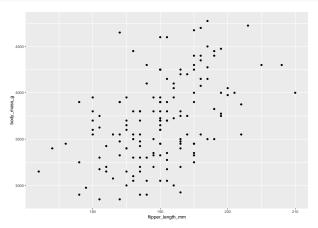
ggplot(data = pdata, aes(x = flipper_length_mm, y = body_mass_g))



Add the points

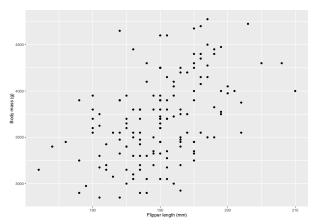
Add layers with ggplot using the +

```
ggplot(data = pdata, aes(x = flipper_length_mm, y = body_mass_g)) +
    geom_point()
```



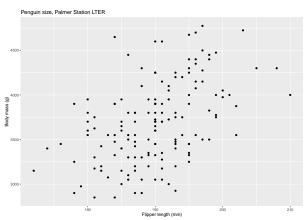
Tidy up labels

```
ggplot(data = pdata, aes(x = flipper_length_mm, y = body_mass_g)) +
   geom_point() +
   xlab("Flipper length (mm)") +
   ylab("Body mass (g)")
```



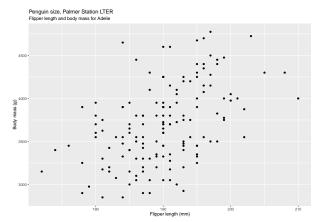
Add a title

```
ggplot(data = pdata, aes(x = flipper_length_mm, y = body_mass_g)) +
geom_point() +
xlab("Flipper length (mm)") +
ylab("Body mass (g)") +
labs(title = "Penguin size, Palmer Station LTER")
```



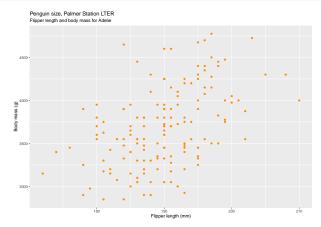
Subtitle

```
ggplot(data = pdata, aes(x = flipper_length_mm, y = body_mass_g)) +
geom_point() +
xlab("Flipper length (mm)") +
ylab("Body mass (g)") +
labs(title = "Penguin size, Palmer Station LTER",
    subtitle = "Flipper length and body mass for Adelie")
```

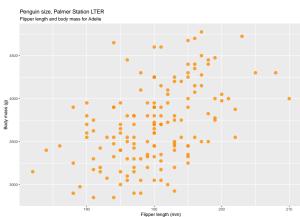


Change color of points

To see all colors, type colors()



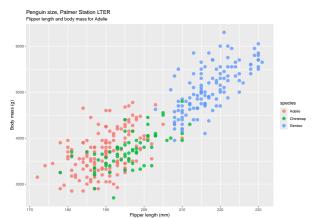
Size of points



Coloring by group

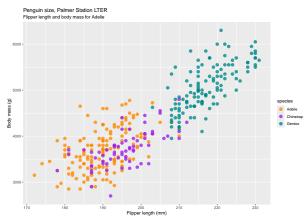
Now use the whole data penguins. Specifying the color in aes() because it **depends on the data**.

```
ggplot(data = penguins, aes(x = flipper_length_mm, y = body_mass_g, color = species)) +
geom_point(size = 3, alpha = 0.8) +
xlab("Flipper length (mm)") +
ylab("Body mass (g)") +
labs(title = "Penguin size, Palmer Station LTER",
    subtitle = "Flipper length and body mass for Adelie")
```

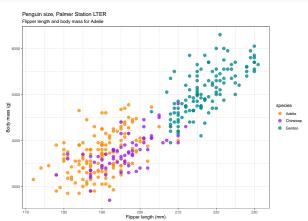


Change the color manually by group

Specifying the color in aes() because it **depends on the data**.



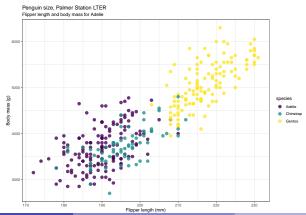
Change the theme



Change the color scheme

Optional: try ggsci functions.

```
ggplot(data = penguins, aes(x = flipper_length_mm, y = body_mass_g, color = species)) +
geom_point(size = 3, alpha = 0.8) +
xlab("Flipper length (mm)") +
ylab("Body mass (g)") +
labs(title = "Penguin size, Palmer Station LTER",
    subtitle = "Flipper length and body mass for Adelie") +
theme_bw() +
scale_color_viridis_d()
```

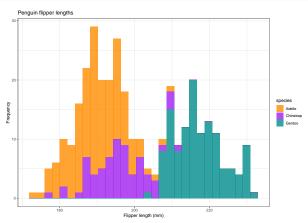


Other plot types?

Other types are also available, e.g. histograms, bar charts, box plots, line graphs and scatter plots.

Histograms

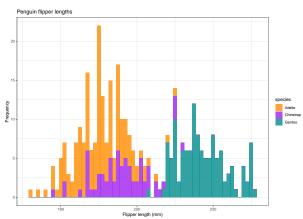
```
ggplot(data = penguins, aes(x = flipper_length_mm, fill = species)) +
geom_histogram(alpha = 0.8) +
scale_fill_manual(values = c("darkorange","purple","cyan4")) +
xlab("Flipper length (mm)") +
ylab("Frequency") +
labs(title = "Penguin flipper lengths") +
theme_bw()
```



Histogram

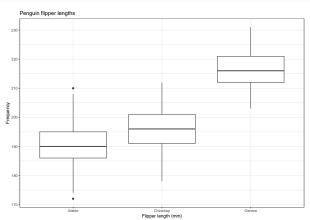
Try binwidth = and bins ==

```
ggplot(data = penguins, aes(x = flipper_length_mm, fill = species)) +
geom_histogram(binwidth = 1, alpha = 0.8) +
scale_fill_manual(values = c("darkorange","purple","cyan4")) +
xlab("Flipper length (mm)") +
ylab("Frequency") +
labs(title = "Penguin flipper lengths") +
theme_bw()
```



Boxplots

```
ggplot(data = penguins, aes(x = species, y = flipper_length_mm)) +
geom_boxplot(show.legend = FALSE) +
xlab("Flipper length (mm)") +
ylab("Frequency") +
labs(title = "Penguin flipper lengths") +
theme_bw()
```



Barcharts

First, try to summarize the penguin data by species and returns the proportion of each penguin types.

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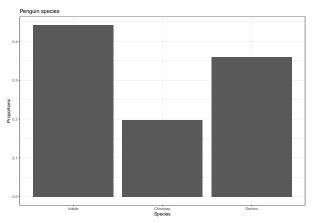
```
gplot <- penguins %>%
  group_by(species) %>%
  tally() %>%
  mutate(prop = n / sum(n))
gplot

## # A tibble: 3 x 3
## species n prop
## <fct> <int> <dbl>
## 1 Adelie 152 0.442
## 2 Chinstrap 68 0.198
```

3 Gentoo 124 0.360

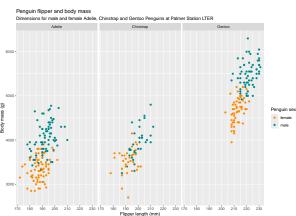
Barcharts

```
ggplot(data = gplot, aes(x = species, y = prop)) +
geom_bar(stat = "identity") +
xlab("Species") +
ylab("Proportions") +
labs(title = "Penguin species") +
theme_bw()
```



Faceting

```
ggplot(penguins, aes(x = flipper_length_mm, y = body_mass_g)) +
geom_point(aes(color = sex)) +
scale_color_manual(values = c("darkorange","cyan4"), na.translate = FALSE) +
labs(title = "Penguin flipper and body mass",
    subtitle = "Dimensions for male and female Adelie, Chinstrap and Gentoo Penguins at Palmer Station LTER"
    x = "Flipper length (mm)",
    y = "Body mass (g)",
    color = "Penguin sex") +
facet_wrap(-species)
```



Git + Github

What is Git?

- A control system to manage projects
- Good for tracking history

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- A control system to manage projects
- Good for tracking history

What about Github?

- Cloud-based service for managing Git repositories
- Useful for teamwork
- Just like "Dropbox"

Why Git + Github?

- You can undo anything
- You won't need to keep undo-ing things (merge/load difference)
- You can identify exactly when and where changes were made
- Teamwork

Some github terminology

- User: A Github account for you (e.g., jules32).
- Organization: The Github account for one or more user (e.g., datacarpentry).
- Repository: A folder within the organization that includes files dedicated to a project.
- Local Github: Copies of Github files located your computer.
- Remote Github: Github files located on the https://github.com website.

Basic Git commands and workflow

When you are working on a your local machine you typically get started by:

git clone: Cloning a remote repository to work on locally. This is a
way to work with an ongoing project or edit someone else's project that
is available remotely (aka on GitHub).

From there, the typical workflow involves:

- git add: Adding files to your repo
- git commit: Commiting changes you have made
- git push: Pushing changes to a remote repository (aka GitHub)

For a collaborative project, or work between desktop and personal laptop, you would use the following first before git add

git pull: Pulling changes from a remote repository

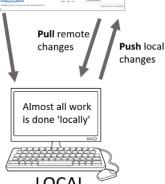
Illustration diagram

REMOTE

(aka Github website)



Clone (i.e., copy) repository to your computer (a one time event)



Let's Git

Download Github and set up your github profile.

- Github Desktop, a GUI for using Github [link]
- (Optional) Learn how to use command line for Github management
 - Command line tutorial [link]

Resources

This tutorial is based on

- Monica Alexander's ggplot tutorial [link]
- Jesse Gronsbell's Github tutorial [link]

Other resources:

https://kbroman.org/github_tutorial/pages/resources.html