# Palmer Penguins Report

#### Ebraheem Amashah

### 1/12/2022

### Setting up my environment

Notes: Setting up the R environment by loading "tidyverse" and "palmerpenguins" packages

```
library(tidyverse)
## -- Attaching packages -----
                              ----- tidyverse 1.3.1 --
## v ggplot2 3.3.5
                  v purrr
                          0.3.4
## v tibble 3.1.2
                  v dplyr
                          1.0.7
## v tidyr 1.1.4 v stringr 1.4.0
## v readr
         2.0.2
                  v forcats 0.5.1
                                     ## -- Conflicts -----
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                masks stats::lag()
```

### Exploratory Analysis

library(palmerpenguins)

Notes: I find it helpful to go through column names and some values before manipulating the data. I find this saves me a lot of time and confusion down the line.

#### colnames (penguins)

```
## [1] "species" "island" "bill_length_mm"
## [4] "bill_depth_mm" "flipper_length_mm" "body_mass_g"
## [7] "sex" "year"
```

#### head(penguins)

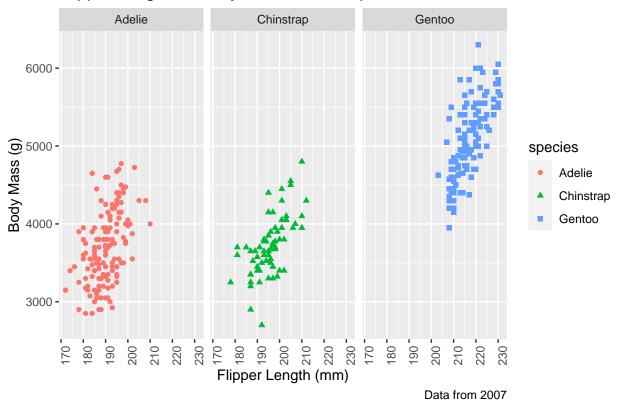
```
## # A tibble: 6 x 8
     species island bill_length_mm bill_depth_mm flipper_length_~ body_mass_g sex
     <fct> <fct>
                             <dbl>
                                           <dbl>
                                                            <int>
                                                                        <int> <fct>
                              39.1
                                            18.7
                                                                         3750 male
## 1 Adelie Torge~
                                                              181
## 2 Adelie Torge~
                              39.5
                                            17.4
                                                              186
                                                                         3800 fema~
## 3 Adelie Torge~
                              40.3
                                            18
                                                                         3250 fema~
                                                              195
## 4 Adelie Torge~
                             NA
                                            NA
                                                               NA
                                                                           NA <NA>
                                                                         3450 fema~
## 5 Adelie Torge~
                              36.7
                                            19.3
                                                              193
## 6 Adelie Torge~
                             39.3
                                            20.6
                                                              190
                                                                         3650 male
## # ... with 1 more variable: year <int>
```

#### Visualization

Notes: I decided to examine the relationship between *body mass* and *flipper length* across the different *species*. We have some null values in the *sex* column that we can get rid of.

```
penguins %>%
  drop_na(sex) %>%
  ggplot(mapping = aes(x = flipper_length_mm, y = body_mass_g,
  color = species, shape = species))+
  geom_point() + facet_wrap(~species) +
  labs(title = "Flipper Lengths & Body Mass Across Species", x = "Flipper Length (mm)", y = "Body Mass theme(axis.text.x = element_text(angle = 90))
```

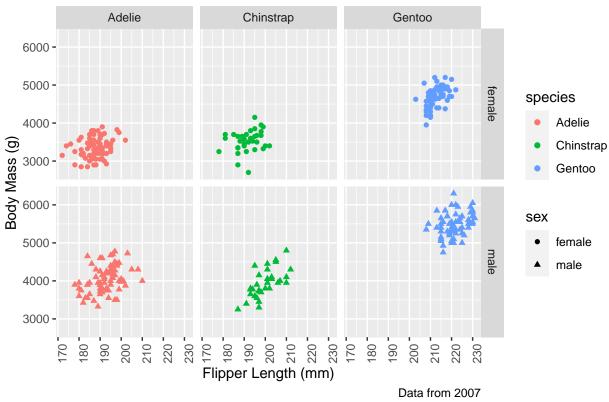
## Flipper Lengths & Body Mass Across Species



Next, I decided to explore the impact of adding sex in the equation. It could be insightful to make comparisons across different sexes along with species.

```
penguins %>%
  drop_na(sex) %>%
  ggplot(mapping = aes(x= flipper_length_mm, y = body_mass_g,
  color = species, shape = sex))+
  geom_point() + facet_grid(sex~species) +
  labs(title = "Flipper Lengths & Body Mass Across Sex & Species",
  x = "Flipper Length (mm)", y = "Body Mass (g)", caption = "Data from 2007") +
  theme(axis.text.x = element_text(angle = 90))
```





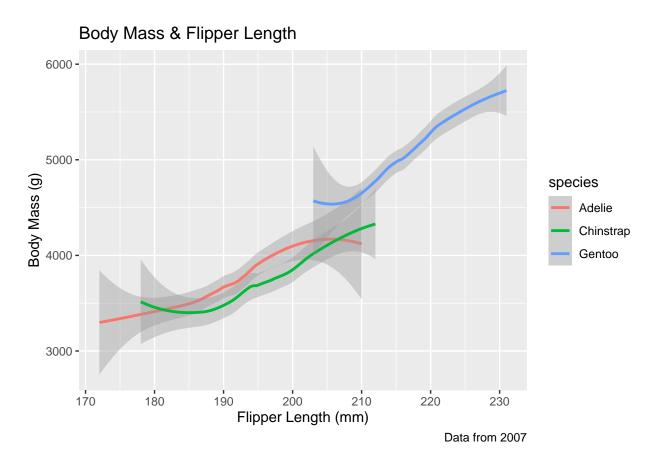
#### Observations and Findings

So far we can see that the male Gentoo penguins are the largest, both in terms of body mass and flipper lengths and the Adelie species in the smallest. We can use a quick statistical analysis code to confirm.

```
penguins %>%
  drop_na() %>%
  group_by(species) %>%
  summarize(mean_body_mass = mean(body_mass_g),
  mean_flipper_length = mean(flipper_length_mm))
## # A tibble: 3 x 3
##
     species
               mean_body_mass mean_flipper_length
     <fct>
                         <dbl>
##
                                              <dbl>
## 1 Adelie
                         3706.
                                               190.
  2 Chinstrap
                         3733.
                                               196.
## 3 Gentoo
                         5092.
                                               217.
```

We can now be assured our initial observations are correct. The largest *species* by both *flipper length* and *body mass* is the Gentoos and the smallest *species* are the Adelie penguins. but is it safe to assume a positive correlation between *body mass* and *flipper length*?

- ## Warning: Ignoring unknown aesthetics: line\_type
- ## 'geom\_smooth()' using method = 'loess' and formula 'y ~ x'
- ## Warning: Removed 2 rows containing non-finite values (stat\_smooth).



This graph clear shows the positive relationship between body mass and flipper length present across all species.