Assignment 3

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1. Exploratory data analysis

```
library (tidyverse)
## -- Attaching core tidyverse packages -
verse 2.0.0 ——
## dplyr 1.1.3
                          √ readr
                                      2.1.4
## ✓ forcats 1.0.0

√ stringr

                                      1.5.0
## J ggplot2 3.4.3
                         √ tibble
                                      3. 2. 1
## / lubridate 1.9.3
                          ✓ tidyr
                                     1.3.0
               1.0.2
## √ purrr
## —— Conflicts ——
--- tidyverse_conflicts() ---
## X dplyr::filter() masks stats::filter()
## X dplyr::lag()
                     masks stats::lag()
## | Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to beco
me errors
```

```
library(Stat2Data)
data("Hawks")
```

1.1 (Q1)

```
# Creating the HawksTail vector
HawksTail <- Hawks$Tail

# Display the first few elements of the vector
head(HawksTail)
```

```
## [1] 219 221 235 220 157 230
```

```
# Output: [1] 219 221 235 220 157 230

# Compute the sample mean and sample median
sample_mean <- mean(HawksTail, na.rm = TRUE) # using na.rm = TRUE to handle any NA values
sample_median <- median(HawksTail, na.rm = TRUE)

# Display the sample mean and sample median
sample_mean</pre>
```

```
## [1] 198.8315
```

```
sample_median
```

```
## [1] 214
```

1.2 (Q1)

```
# Using summarise() to compute mean, trimmed mean, and median for Wing and Weight columns
result <- Hawks %>%
summarise(
    Wing_mean = mean(Wing, na.rm = TRUE),
    Wing_t_mean = mean(Wing, trim = 0.5, na.rm = TRUE), # trimmed mean with q=0.5
    Wing_med = median(Wing, na.rm = TRUE),
    Weight_mean = mean(Weight, na.rm = TRUE),
    Weight_t_mean = mean(Weight, trim = 0.5, na.rm = TRUE), # trimmed mean with q=0.5
    Weight_med = median(Weight, na.rm = TRUE)
)
# Display the result
print(result)
```

```
## Wing_mean Wing_t_mean Wing_med Weight_mean Weight_t_mean Weight_med ## 1 315.6375 370 370 772.0802 970 970
```

1.2 (Q2)

```
# Grouping by Species and then using summarise() to compute mean, trimmed mean, and median
grouped_result <- Hawks %>%
  group_by(Species) %>%
  summarise(
    Wing_mean = mean(Wing, na.rm = TRUE),
    Wing_t_mean = mean(Wing, trim = 0.5, na.rm = TRUE),  # trimmed mean with q=0.5
    Wing_med = median(Wing, na.rm = TRUE),
    Weight_mean = mean(Weight, na.rm = TRUE),
    Weight_t_mean = mean(Weight, trim = 0.5, na.rm = TRUE),  # trimmed mean with q=0.5
    Weight_med = median(Weight, na.rm = TRUE)
)

# Display the grouped result
print(grouped_result)
```

```
## # A tibble: 3 	imes 7
     Species Wing mean Wing t mean Wing med Weight mean Weight t mean Weight med
##
     <fct>
                  <db1>
                               <db1>
                                         <db1>
                                                      <db1>
                                                                     <db1>
                                                                                 <db1>
## 1 CH
                   244.
                                 240
                                           240
                                                       420.
                                                                      378.
                                                                                  378.
## 2 RT
                   383.
                                 384
                                           384
                                                      1094.
                                                                     1070
                                                                                 1070
## 3 SS
                   185.
                                 191
                                           191
                                                       148.
                                                                      155
                                                                                  155
```

. . . .

1.3 (Q1)

```
a <- 2
b <- 3

transformed_mean <- mean(HawksTail * a + b)
calculated_mean <- a * mean(HawksTail) + b

# Compare the two means:
transformed_mean</pre>
```

[1] 400.663

 ${\tt calculated_mean}$

[1] 400.663

1,3 (Q2)

```
transformed_variance <- var(HawksTail * a + b)
calculated_variance <- a^2 * var(HawksTail)

transformed_sd <- sd(HawksTail * a + b)
calculated_sd <- a * sd(HawksTail)

# Compare the variance and standard deviation:
transformed_variance</pre>
```

```
## [1] 5424.147
```

```
calculated_variance
 ## [1] 5424.147
 transformed sd
 ## [1] 73.64881
 calculated_sd
 ## [1] 73.64881
1.4
 hal <- Hawks $ Hallux # Extract the vector of hallux lengths
 hal <-hal[!is.na(hal)] # Remove any nans
 \verb"outlier_val<-100"
 num outliers<-10
 corrupted_hal<-c(hal, rep(outlier_val, times=num_outliers))</pre>
 mean(hal)
 ## [1] 26.41086
 mean(corrupted_hal)
 ## [1] 27.21776
 num_outliers_vect \leftarrow seq(0, 1000)
 means_vect <- c()</pre>
 for(num_outliers in num_outliers_vect) {
 corrupted_hal <- c(hal, rep(outlier_val, times=num_outliers))</pre>
 means_vect <- c(means_vect, mean(corrupted_hal))</pre>
1.4 (Q1)
 num_outliers_vect <- seq(0,1000)</pre>
 medians vect <- c()
```

```
num_outliers_vect <- seq(0,1000)
medians_vect <- c()

for(num_outliers in num_outliers_vect) {
   corrupted_hal <- c(hal, rep(outlier_val, times=num_outliers))
   medians_vect <- c(medians_vect, median(corrupted_hal))
}</pre>
```

1.4 (Q2)

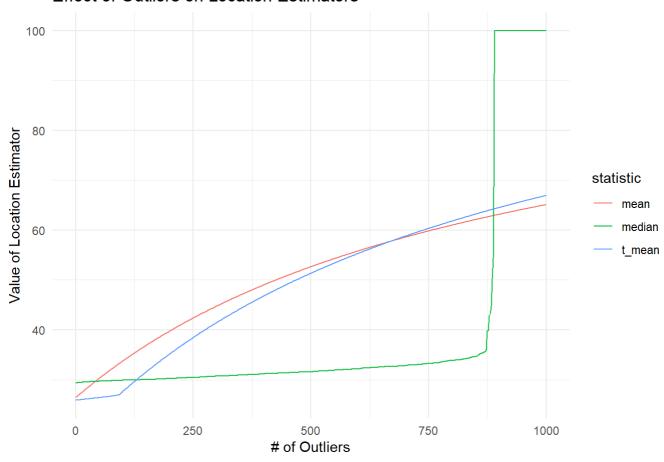
```
num_outliers_vect <- seq(0,1000)
t_means_vect <- c()

for(num_outliers in num_outliers_vect) {
   corrupted_hal <- c(hal, rep(outlier_val, times=num_outliers))
   t_means_vect <- c(t_means_vect, mean(corrupted_hal, trim = 0.1))
}</pre>
```

1.4 (Q3)

```
# Combining into a dataframe
df means medians <- data.frame(
  num outliers=num outliers vect,
  mean=means_vect,
  t mean=t means vect,
  median=medians_vect
)
# Reshape data for plotting
df_long <- df_means_medians %>%
  pivot_longer(
    cols = -num_outliers,
    names_to = "statistic",
    values_to = "value"
  )
# Plotting
ggplot(df_long, aes(x=num_outliers, y=value, color=statistic)) +
  geom_1ine() +
  labs(title="Effect of Outliers on Location Estimators",
       x="# of Outliers",
       y="Value of Location Estimator") +
  theme_minimal()
```

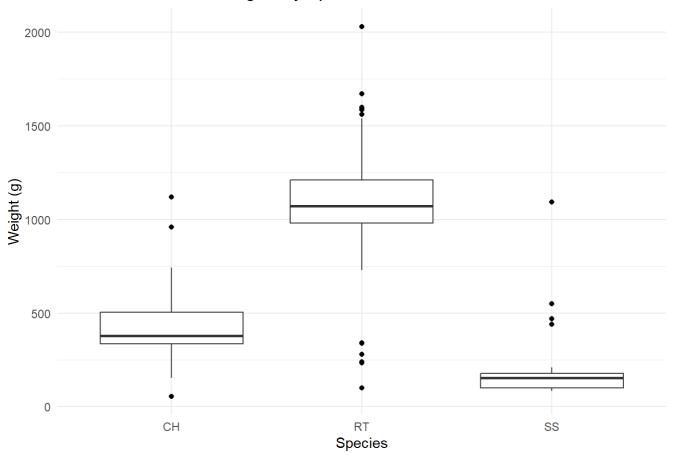
Effect of Outliers on Location Estimators



1.5(Q1)

Warning: Removed 10 rows containing non-finite values (`stat_boxplot()`).

Distribution of Hawk Weights by Species



1.5(Q2)

```
# Grouping by species and computing quantiles
quantiles_df <- Hawks %>%
group_by(Species) %>%
summarise(
   quantile025 = quantile(Weight, 0.25, na.rm = TRUE),
   quantile050 = quantile(Weight, 0.50, na.rm = TRUE),
   quantile075 = quantile(Weight, 0.75, na.rm = TRUE)
)
print(quantiles_df)
```

```
## # A tibble: 3 \times 4
##
     Species quantile025 quantile050 quantile075
##
                    <db1>
                                 <db1>
                                              <db1>
## 1 CH
                       335
                                  378.
                                                505
## 2 RT
                      980
                                 1070
                                              1210
## 3 SS
                      100
                                  155
                                                178.
```

quantile025 corresponds to the lower hinge (the bottom of the box) of the boxplot. quantile050 corresponds to the median (the line inside the box) of the boxplot. quantile075 corresponds to the upper hinge (the top of the box) of the boxplot. ## 1.5(Q3)

```
num_outliers <- function(sample_vector) {
    # Removing NA values
    sample_vector <- sample_vector[!is.na(sample_vector)]

# Calculating the quantiles
    q25 <- quantile(sample_vector, 0.25)
    q75 <- quantile(sample_vector, 0.75)

# Calculating the Interquartile Range (IQR)
    IQR <- q75 - q25

# Finding outliers based on the provided conditions
lower_bound <- q25 - 1.5 * IQR
    upper_bound <- q75 + 1.5 * IQR

outliers <- sample_vector[sample_vector < lower_bound | sample_vector > upper_bound]
    return(length(outliers))
}

# Testing the function
num_outliers( c(0, 40, 60, 185))
```

[1] 1

1.5(Q4)

```
outliers_by_species <- Hawks %>%
  group_by(Species) %>%
  summarise(
    num_of_outliers = num_outliers(Weight)
)
print(outliers_by_species)
```

1.6(Q1)

```
# Compute covariance
cov_weight_wing <- cov(Hawks$Weight, Hawks$Wing, use = "complete.obs")

# Compute correlation
cor_weight_wing <- cor(Hawks$Weight, Hawks$Wing, use = "complete.obs")

cat("Covariance between Weight and Wing:", cov_weight_wing, "\n")

## Covariance between Weight and Wing: 41174.39

cat("Correlation between Weight and Wing:", cor_weight_wing, "\n")

## Correlation between Weight and Wing: 0.9348575
```

1.6(Q2)

```
# Assuming you have a dataframe Hawks with columns Weight and Wing

a <- 2.4
b <- 7.1
c <- -1
d <- 3

# Creating the transformed variables
Hawks$Weight_transformed <- a * Hawks$Weight + b
Hawks$Wing_transformed <- c * Hawks$Wing + d

# Computing covariance and correlation for transformed variables
cov_transformed <- cov(Hawks$Weight_transformed, Hawks$Wing_transformed, use = "complete.obs")
cor_transformed <- cov(Hawks$Weight_transformed, Hawks$Wing_transformed, use = "complete.obs")

# Displaying the results
cat("Covariance of transformed variables: ", cov_transformed, "\n")
```

```
cat("Correlation of transformed variables:", cor_transformed, "\n")

## Correlation of transformed variables: -0.9348575
```

2.1(Q1)

- 1. Random Experiment: A random experiment is an experiment or a process for which the outcome cannot be predicted with certainty.
- 2. Sample Space: The sample space, often denoted as S or Ω , refers to the set of all possible outcomes of a random experiment. It encompasses every conceivable result for the given experiment

3. Event: An event is any subset of the sample space. It represents a specific set of outcomes of a random experiment that we might be interested in.

2.1 (Q2)

2.2 (Q1)

2.2 (Q2)

2.2 (Q2)

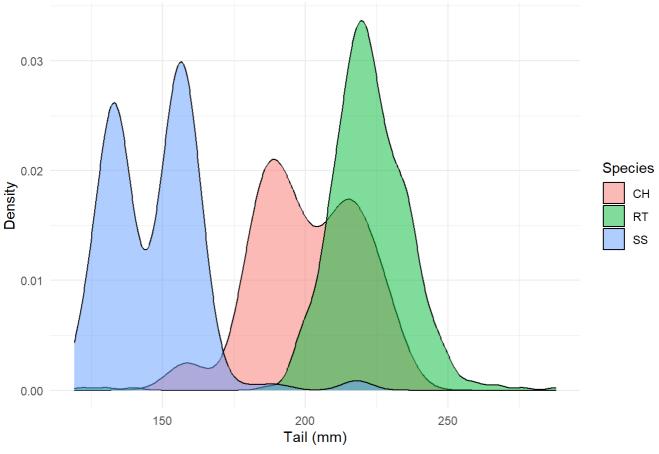
2.2 (Q3)

2.2 (Q4)

2.2 (Q5)

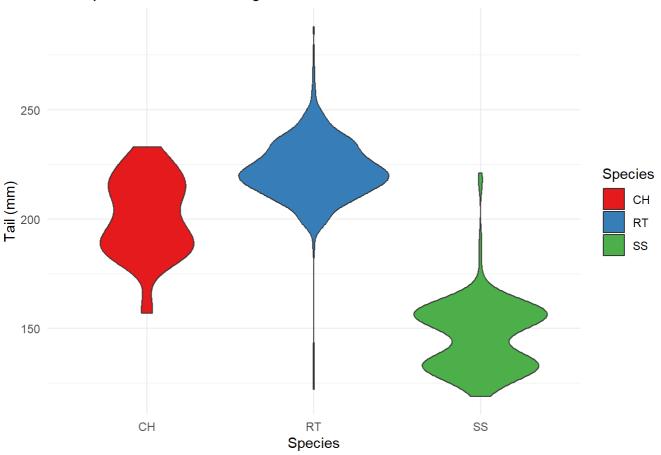
3 (Q1)

Density plot of Hawk Tail Lengths by Species



3 (Q2)

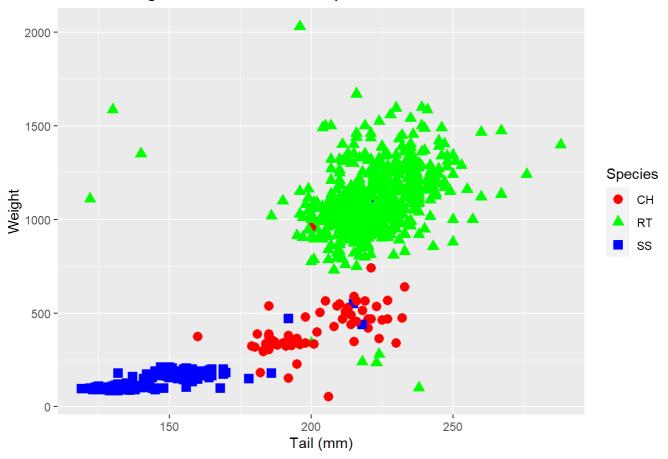
Violin plot of Hawk Tail Lengths



3 (Q3)

Warning: Removed 10 rows containing missing values (`geom point()`).

Tail vs. Weight for different Hawks species



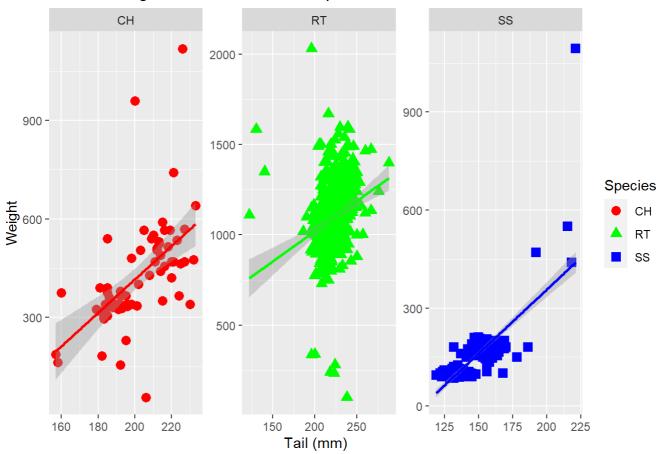
3 (Q4)

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

```
## Warning: Removed 10 rows containing non-finite values (`stat_smooth()`).
```

```
## Warning: Removed 10 rows containing missing values (`geom_point()`).
```

Tail vs. Weight for different Hawks species



..... ## 3 (Q5)

```
# Find the heaviest hawk
heaviest_hawk <- Hawks %>%
  filter(Weight == max(Weight, na.rm = TRUE)) %>%
  select(Tail, Weight, Species) %>%
  top_n(1)
```

Selecting by Species

```
tail_heaviest <- heaviest_hawk$Tail</pre>
weight_heaviest <- heaviest_hawk$Weight</pre>
# Define the custom shapes and colors for species
hawks\_shapes \leftarrow c (CH = 16, RT = 17, SS = 15)  # 16: circle, 17: triangle, 15: square
hawks_colors <- c(CH = "red", RT = "green", SS = "blue")
# Plot
ggplot(Hawks, aes(x = Tail, y = Weight, shape = Species, color = Species)) +
  geom point(aes(shape = Species, color = Species), size = 3) +
  scale shape manual(values = hawks shapes) +
  scale color manual(values = hawks colors) +
  labs(title = "Tail vs. Weight for different Hawks species",
       x = "Tail (mm)",
       y = "Weight",
       shape = "Species",
       color = "Species") +
  geom\_segment(aes(x = tail\_heaviest, xend = tail\_heaviest + 10,
                   y = weight heaviest, yend = weight heaviest + 100),
               arrow = arrow(type = "closed", length = unit(0.2, "inches")), color = "black") +
  annotate ("text", x = tail_heaviest + 25, y = weight_heaviest + 150,
           label = "heaviest hawk", color = "black")
```

Warning: Removed 10 rows containing missing values (`geom_point()`).

Tail vs. Weight for different Hawks species

