

Lab2

Exercises

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Contents

Exercise 1 - Create a new mean and sd function	1
A	1
B	2
C	2
D	2
E	3
Exercise 2 - Table of frequencies	4
A	4
B	5
C	6
D	6
Exercise 3 - Histogram, Boxplot and quartiles	7
A	7
B	8
C	8
D	8
E	8
Exercise 4 - Multiple boxplots from scratch	9
A	9
B	9
C	9
Exercise 5 - Exploratory analysis of data	9
A	9
B	9
C	9
D	9
E	9
F	10

Exercise 1 - Create a new mean and sd function

A

Create the Lab2 project. Use the same structure used for Lab1:

- scripts,
- plots,

- data

B

Install the palmerpenguins package, load the penguins dataset or, alternatively, download the .RData object from moodle and import it after placing it inside the data directory of the project (hint: use the load() function).

```
library(palmerpenguins)
data(penguins)
```

C

Compute the mean, the standard deviation, and the median for the numeric variables of the dataset.

```
# Means
cat("Means: \n")
## Means:
colMeans(penguins[, c(3:6, 8)], na.rm = TRUE)
##      bill_length_mm      bill_depth_mm flipper_length_mm      body_mass_g
##           43.92193           17.15117           200.91520           4201.75439
##              year
##           2008.02907
cat("\n")

# Medians
cat("Medians: \n")
## Medians:
sapply(penguins[, c(3:6, 8)], median, na.rm = TRUE)
##      bill_length_mm      bill_depth_mm flipper_length_mm      body_mass_g
##           44.45           17.30           197.00           4050.00
##              year
##           2008.00
cat("\n")

# Standard deviations
cat("Standard deviations: \n")
## Standard deviations:
sapply(penguins[, c(3:6, 8)], sd, na.rm = TRUE)
##      bill_length_mm      bill_depth_mm flipper_length_mm      body_mass_g
##           5.4595837           1.9747932           14.0617137           801.9545357
##              year
##           0.8183559
cat("\n")
```

D

Create a function called stat_auto that simultaneously returns both the mean and the standard deviation of a given vector (hint: return an object of type list or simply a vector). Then try it on the same numeric variables in C. to check the results (hint: if you obtain NA maybe you forgot to remove NA terms in the vector).

```
stat_auto <- function(vec, na.rm = FALSE) {
  if (na.rm) {
    mean <- mean(vec, na.rm = TRUE)
```

```

    sd <- sd(vec, na.rm = TRUE)

    return(list("mean" = mean, "sd" = sd))
  }

mean <- mean(vec)
sd <- sd(vec)

return(list("mean" = mean, "sd" = sd))
}

sapply(penguins[, c(3:6, 8)], stat_auto, na.rm = TRUE)

##      bill_length_mm bill_depth_mm flipper_length_mm body_mass_g
## mean 43.92193      17.15117      200.9152          4201.754
## sd   5.459584      1.974793      14.06171          801.9545
##      year
## mean 2008.029
## sd   0.8183559

```

E

Create a function called `stat_manual` that simultaneously returns both the mean and the standard deviation of a given vector without using the `mean()` and the `sd()` functions (hint: you can use `length()`, `sum()`, and `na.omit()` functions). Then try it on the same numeric variables in C. to check the results.

```

stat_manual <- function(vec, na.rm = FALSE) {
  if (na.rm) {
    sum <- sum(vec, na.rm = TRUE)
    mean <- sum / na.omit(length(vec))

    sum <- sum((vec - mean)^2, na.rm = TRUE)
    denom <- na.omit(length(vec)) - 1
    varianza <- sum / denom

    sd <- sqrt(varianza)

    return(list("mean" = mean, "sd" = sd))
  }

  sum <- sum(vec)
  mean <- sum / length(vec)

  sum <- sum((vec - mean)^2)
  denom <- length(vec) - 1
  varianza <- sum / denom

  sd <- sqrt(varianza)

  return(list("mean" = mean, "sd" = sd))
}

sapply(penguins[, c(3:6, 8)], stat_manual, na.rm = TRUE)

##      bill_length_mm bill_depth_mm flipper_length_mm body_mass_g

```

```
## mean 43.66657      17.05145      199.7471      4177.326
## sd   5.449612      1.971543      14.06909      799.985
##      year
## mean 2008.029
## sd   0.8183559
```

Exercise 2 - Table of frequencies

A

In the penguins dataset, transform a numeric variable to a categorical one by aggregating values into classes. Consider the flipper length variable and create 10mm wide classes using the `cut()` function (hint: use the `range()` function to determine the min and max values of the variable, then define a sequence for the cuts).

```
r <- range(penguins$flipper_length_mm, na.rm = TRUE)
splits <- seq(r[1], r[2], 10)
splits <- append(splits, r[2])

classes <- cut(penguins$flipper_length_mm, splits, ordered_result = TRUE)

cat("Splits: ", splits, "\n")
```

```
## Splits: 172 182 192 202 212 222 231
```

```
cat("Classes: \n")
```

```
## Classes:
```

```
classes
```

```
## [1] (172,182] (182,192] (192,202] <NA>      (192,202] (182,192]
## [7] (172,182] (192,202] (192,202] (182,192] (182,192] (172,182]
## [13] (172,182] (182,192] (192,202] (182,192] (192,202] (192,202]
## [19] (182,192] (192,202] (172,182] (172,182] (182,192] (182,192]
## [25] (172,182] (182,192] (182,192] (182,192] <NA>      (172,182]
## [31] (172,182] (172,182] (182,192] (182,192] (192,202] (192,202]
## [37] (182,192] (172,182] (172,182] (182,192] (172,182] (192,202]
## [43] (182,192] (192,202] (182,192] (182,192] (172,182] (172,182]
## [49] (182,192] (182,192] (182,192] (182,192] (182,192] (192,202]
## [55] (182,192] (182,192] (182,192] (192,202] (172,182] (192,202]
## [61] (182,192] (192,202] (182,192] (182,192] (182,192] (182,192]
## [67] (192,202] (182,192] (182,192] (192,202] (182,192] (182,192]
## [73] (192,202] (192,202] (182,192] (192,202] (182,192] (182,192]
## [79] (182,192] (192,202] (182,192] (192,202] (182,192] (192,202]
## [85] (182,192] (192,202] (182,192] (182,192] (182,192] (182,192]
## [91] (192,202] (202,212] (182,192] (182,192] (182,192] (202,212]
## [97] (182,192] (192,202] (172,182] (182,192] (182,192] (202,212]
## [103] (182,192] (182,192] (192,202] (182,192] (192,202] (182,192]
## [109] (172,182] (192,202] (192,202] (182,192] (192,202] (192,202]
## [115] (182,192] (192,202] (182,192] (192,202] (182,192] (182,192]
## [121] (182,192] (192,202] (172,182] (192,202] (182,192] (192,202]
## [127] (182,192] (192,202] (182,192] (202,212] (182,192] (192,202]
## [133] (192,202] (192,202] (182,192] (182,192] (182,192] (192,202]
## [139] (182,192] (192,202] (192,202] (182,192] (182,192] (182,192]
## [145] (182,192] (182,192] (182,192] (182,192] (192,202] (192,202]
## [151] (182,192] (192,202] (202,212] (222,231] (202,212] (212,222]
```

```
## [157] (212,222] (202,212] (202,212] (212,222] (202,212] (212,222]
## [163] (212,222] (212,222] (212,222] (212,222] (202,212] (212,222]
## [169] (202,212] (212,222] (202,212] (212,222] (212,222] (212,222]
## [175] (212,222] (212,222] (212,222] (212,222] (212,222] (212,222]
## [181] (202,212] (212,222] (212,222] (202,212] (202,212] (222,231]
## [187] (212,222] (212,222] (212,222] (212,222] (202,212] (202,212]
## [193] (202,212] (222,231] (202,212] (212,222] (212,222] (212,222]
## [199] (202,212] (222,231] (212,222] (212,222] (202,212] (212,222]
## [205] (202,212] (222,231] (212,222] (212,222] (202,212] (212,222]
## [211] (202,212] (222,231] (202,212] (212,222] (212,222] (222,231]
## [217] (212,222] (222,231] (212,222] (222,231] (212,222] (222,231]
## [223] (212,222] (212,222] (212,222] (212,222] (212,222] (222,231]
## [229] (202,212] (212,222] (212,222] (222,231] (202,212] (212,222]
## [235] (202,212] (222,231] (202,212] (222,231] (212,222] (212,222]
## [241] (202,212] (222,231] (212,222] (222,231] (202,212] (222,231]
## [247] (212,222] (222,231] (212,222] (212,222] (202,212] (222,231]
## [253] (212,222] (222,231] (212,222] (222,231] (212,222] (212,222]
## [259] (202,212] (212,222] (202,212] (202,212] (212,222] (222,231]
## [265] (212,222] (222,231] (212,222] (222,231] (212,222] (212,222]
## [271] (212,222] <NA> (212,222] (212,222] (202,212] (212,222]
## [277] (182,192] (192,202] (192,202] (182,192] (192,202] (192,202]
## [283] (172,182] (192,202] (192,202] (192,202] (192,202] (192,202]
## [289] (182,192] (192,202] (182,192] (192,202] (192,202] (172,182]
## [295] (182,192] (192,202] (172,182] (182,192] (182,192] (192,202]
## [301] (192,202] (192,202] (192,202] (192,202] (182,192] (202,212]
## [307] (182,192] (192,202] (182,192] (202,212] (192,202] (192,202]
## [313] (192,202] (202,212] (182,192] (202,212] (202,212] (182,192]
## [319] (192,202] (192,202] (192,202] (192,202] (182,192] (202,212]
## [325] (182,192] (192,202] (192,202] (192,202] (192,202] (202,212]
## [331] (182,192] (192,202] (182,192] (202,212] (192,202] (192,202]
## [337] (202,212] (182,192] (192,202] (202,212] (192,202] (192,202]
## [343] (202,212] (192,202]
## 6 Levels: (172,182] < (182,192] < (192,202] < ... < (222,231]
```

B

Use the `table()` function on the new variable generated by `cut()`. Then transform it into a `data.frame` object. Rename the columns accordingly using the `colnames()` function (hint: the second column correspond to the absolute frequencies).

```
df <- data.frame(table(classes))
colnames(df) <- c("Classes", "AbsoluteFrequencies")
df
```

```
##      Classes AbsoluteFrequencies
## 1 (172,182]             22
## 2 (182,192]             96
## 3 (192,202]             85
## 4 (202,212]             47
## 5 (212,222]             67
## 6 (222,231]             24
```

C

Add the the columns for: relative frequencies, cumulative absolute frequencies, and cumulative relative frequencies.

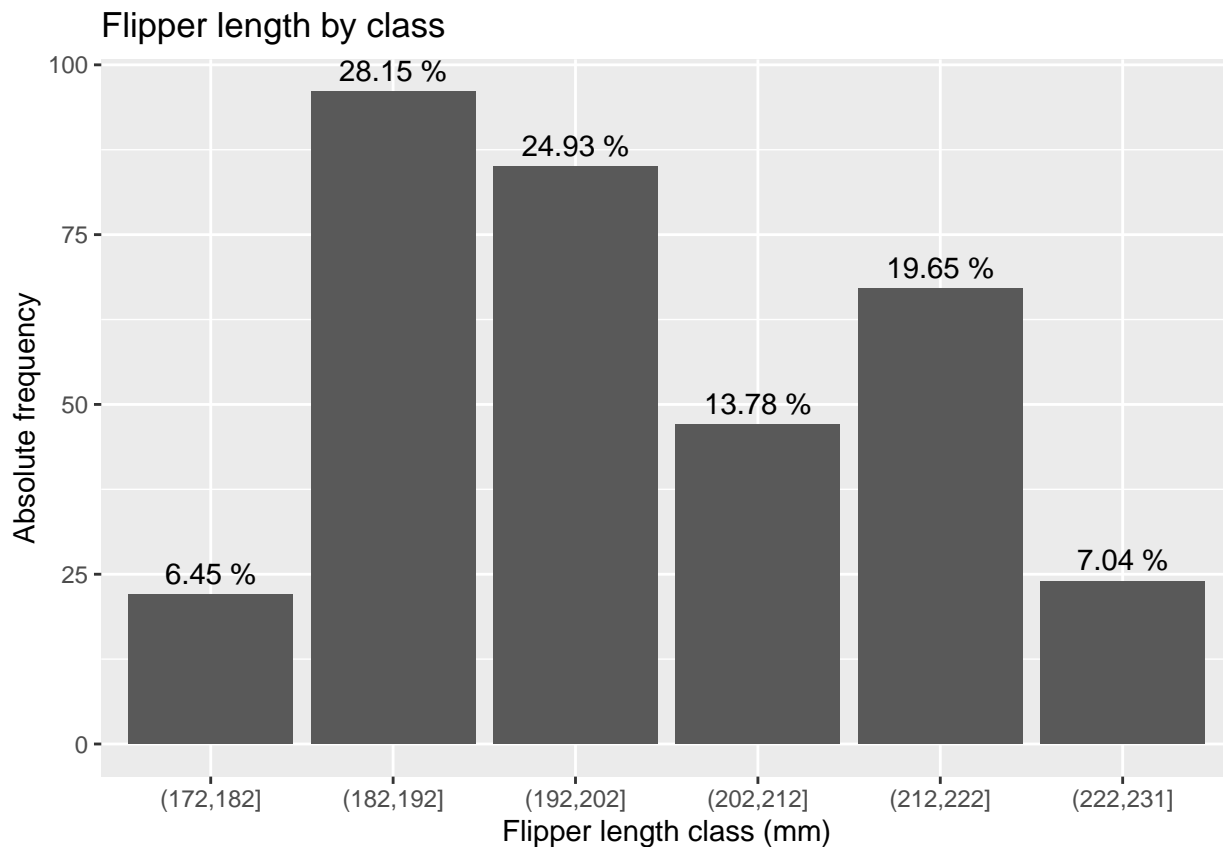
```
df$CumAbsFreq <- cumsum(df$AbsoluteFrequencies)
df$RelativeFrequencies <- df$AbsoluteFrequencies / sum(df$AbsoluteFrequencies)
df$RelAbsFreq <- cumsum(df$RelativeFrequencies)
df
```

```
##      Classes AbsoluteFrequencies CumAbsFreq RelativeFrequencies
## 1 (172,182]                22         22      0.06451613
## 2 (182,192]                96        118      0.28152493
## 3 (192,202]                85        203      0.24926686
## 4 (202,212]                47        250      0.13782991
## 5 (212,222]                67        317      0.19648094
## 6 (222,231]                24        341      0.07038123
## RelAbsFreq
## 1 0.06451613
## 2 0.34604106
## 3 0.59530792
## 4 0.73313783
## 5 0.92961877
## 6 1.00000000
```

D

Use the `geom_col()` function to plot the frequency of each class. Then, using the `geom_text(aes(label = ...))` function, add the relative frequency as a percentage above each column (hint: substitute the ... with the relative frequency values. Use the `round()` function to choose the appropriate number of digits).

```
library(ggplot2)
ggplot(
  data = df,
  aes(
    x = df$Class,
    y = df$AbsoluteFrequency,
  )
) +
  geom_col(aes(y = df$AbsoluteFrequencies)) +
  geom_text(
    aes(
      label = paste(round(df$RelativeFrequencies * 100, 2), "%"),
      y = df$AbsoluteFrequencies,
    ),
    vjust = -0.5,
  ) +
  labs(
    title = "Flipper length by class",
    x = "Flipper length class (mm)",
    y = "Absolute frequency",
  ) +
  theme(legend.position = "bottom")
```



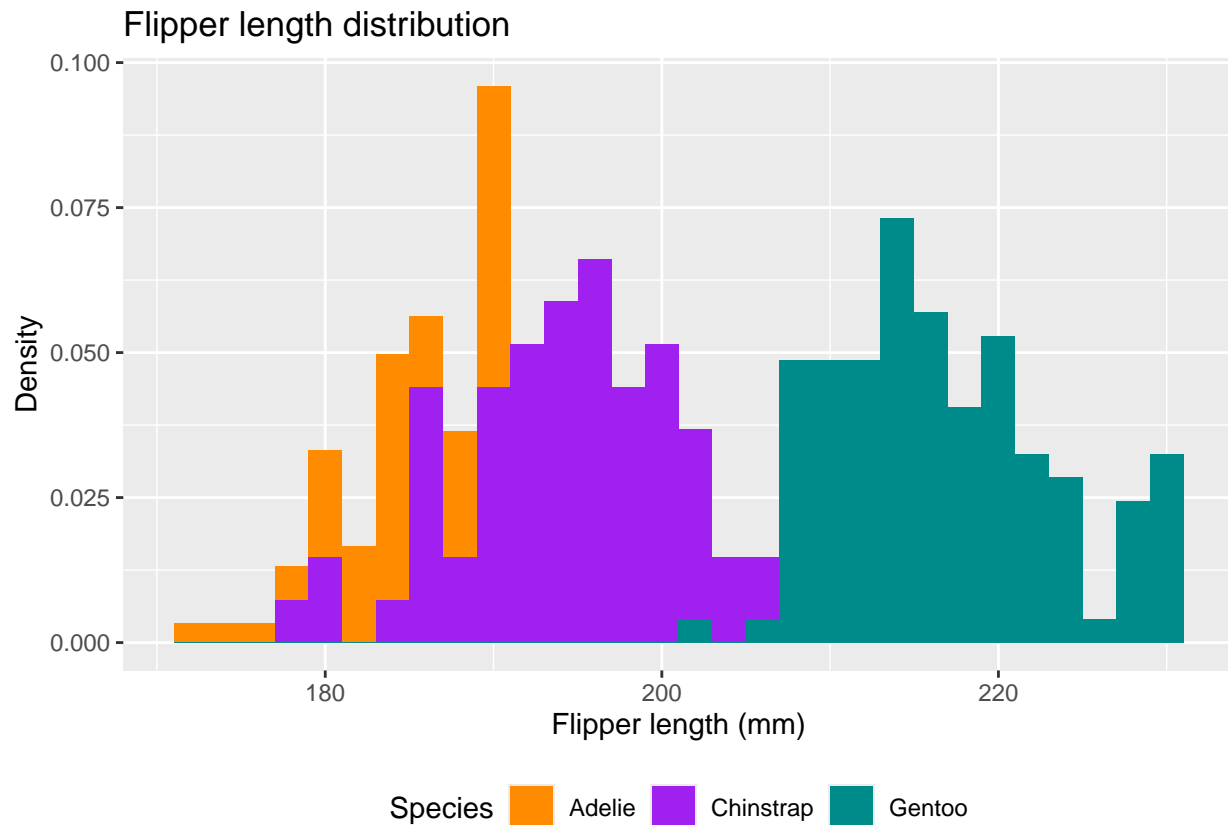
Exercise 3 - Histogram, Boxplot and quartiles

A

Using the `geom_histogram()` function of the `ggplot2` package plot the flipper length distribution coloring each species with a different color (hint: use the `fill` argument of the `aes()` function to fill the histogram area and the `position = "identity"` argument of the `geom_histogram()`). Play with the `binwidth` argument. Try to insert `y = ..density..` in `aes()`. Do you notice any change?

```
ggplot(
  data = penguins,
  aes(
    x = penguins$flipper_length_mm,
    fill = penguins$species,
  )
) +
  scale_fill_manual(values = c("darkorange", "purple", "cyan4")) +
  geom_histogram(
    position = "identity",
    binwidth = 2,
    aes(y = ..density..),
  ) +
  labs(
    title = "Flipper length distribution",
    x = "Flipper length (mm)",
    y = "Density",
    fill = "Species",
  )
```

```
) +  
theme(legend.position = "bottom")
```



B

About the flipper length, for each species of penguins compute the: 1. Sample mean 2. Sample median 3. Sample standard deviation (use a division by $n - 1$) 4. Sample variance

(hint: to choose only a specific species use `penguins[penguins$species == "Gentoo",]`)

C

Using the `geom_boxplot()` function of the `ggplot2` package plot the boxplot for the flipper length variable coloring each species with a different color (hint: use the `color` argument of the `aes()` function).

D

Compute the flipper length quartiles for the "Gentoo" penguins (Q1, Q2, Q3).

E

Calculate the flipper length 40th percentile for the "Adelie" penguins.

Exercise 4 - Multiple boxplots from scratch

A

Generate random data with some structure, and create one data set for each day of the week (hint: use the `for()` cycle, data should have 7 columns). At the end you should obtain a matrix with N rows (N = number of random number to generate each time) and 7 columns (one for each day of the week).

B

Go from a wide to a long data format. You should create a `data.frame` object with exactly two columns. One contains the values created in A., the other contains the corresponding day of the week.

C

Plot the seven boxplots (one for each day of the week) in one graph, horizontally oriented (hint: `coord_flip()` function translates the axes, the “limits” argument of `scale_x_discrete()` allows you to reorder the axis labels).

Exercise 5 - Exploratory analysis of data

Penguins dataset does not contain their weights and flipper lengths only. Many other variables are available. Let's explore it a little more:

A

How many islands are there? And how many penguins are present in each isle? Are the 3 species of penguins living together? (hint: use the `table()` function).

B

Try to use the `geom_bar()` or `geom_col()` functions to graphically represent the population of each island, colored by species (hint: islands in the x-axis, number of penguins in the y-axis).

C

Use a scatter plot to represent flipper length vs. body mass. Color the point according to the “sex” variable. Try to use facets to see if there are differences across species (hint: use `facet_grid(~ species)` function to add facets for species).

D

The numeric variables shows some interesting relationships. Are they correlated? Use the `cor()` and the `corrplot()` functions to study correlations between numeric variables (hint: try to google `corrplot()` to see which package you have to install to use it).

E

Choose a pair of numeric variables, compute the correlation between them without using the `cor()` function (hint: remember the formula).

F

Plot the scatter plot for bill length vs. bill depth. Color the points by species. Use the function `geom_smooth(formula = "y ~ x")` to add a line to represent the linear relationship between the two variables. Then, again, use `geom_smooth(formula = "y ~ x")` colored by species. What are you noticing?