Programming in R: Solution for practical session 6a, Topics in Tidyverse and data analysis, (14/11/2023)

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General information

- The practical session is focus on statistical modeling and Tidyverse and consists of 6 questions in which you are asked to conduct an analysis of a dataset. Dataset is available in R as a data frame as a part of the R package palmerpenguins.
- Your output should consists a PDF file which contains the results and R code.
- Solutions will be available online in BB in a later stage (you will receive an email about this via BB).

R functions

Some of the R functions that are used in this practical session are:

- lm(), aov().
- ggplot(), boxplot()....

Question 1:

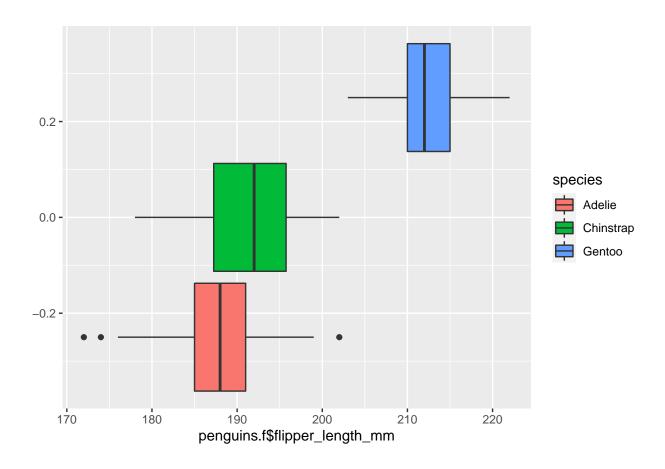
In questions Q1-Q6 we focus on the penguins data which available as a part of the palmerpenguins R package. To assess the data, you need to install the package. More information about the data is available in https://allisonhorst.github.io/palmerpenguins/.For all the questions (Q1-Q5) we conduct a complete case analysis, i.e., all observations should not have missing values. The first 6 lines in the data are shown below

```
library(palmerpenguins)
data("penguins", package = "palmerpenguins")
penguins <- drop_na(penguins)
#dim(penguins)
#table(penguins$sex)
head(penguins)</pre>
```

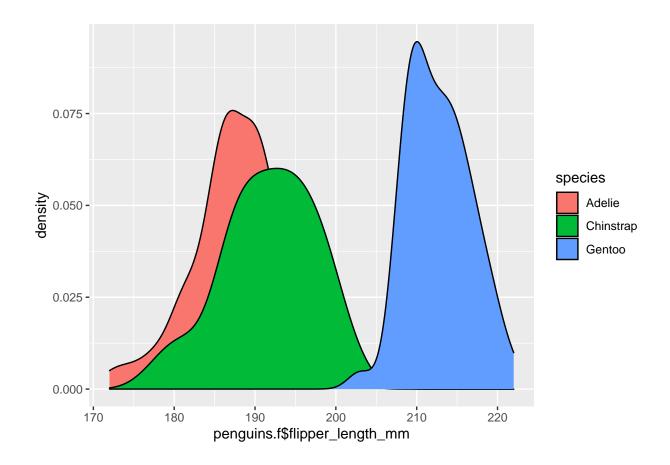
```
## # A tibble: 6 x 8
##
     species island
                        bill_length_mm bill_depth_mm flipper_1~1 body_~2 sex
                                                                                  year
##
     <fct>
             <fct>
                                 <dbl>
                                                <dbl>
                                                                    <int> <fct> <int>
                                                            <int>
## 1 Adelie Torgersen
                                  39.1
                                                18.7
                                                              181
                                                                     3750 male
                                                                                  2007
## 2 Adelie Torgersen
                                  39.5
                                                              186
                                                                     3800 fema~
                                                                                  2007
                                                17.4
                                                                                  2007
## 3 Adelie Torgersen
                                  40.3
                                                18
                                                              195
                                                                     3250 fema~
## 4 Adelie Torgersen
                                  36.7
                                                19.3
                                                              193
                                                                     3450 fema~
                                                                                  2007
## 5 Adelie Torgersen
                                  39.3
                                                20.6
                                                              190
                                                                     3650 male
                                                                                  2007
## 6 Adelie Torgersen
                                  38.9
                                                17.8
                                                              181
                                                                     3625 fema~
                                                                                 2007
## # ... with abbreviated variable names 1: flipper_length_mm, 2: body_mass_g
```

- 1. Create a new data frame contains only the female.
- 2. For the new data frame, calculate the mean, median and standard deviation for the variable flipper_length_mm by species and produce the table below.
- 3. Produce the figures below.

```
penguins.f<-penguins %>% filter(sex=="female")
XXX<-penguins.f %>% group_by(species) %>% summarize(average = mean(flipper_length_mm), median = median(
print(XXX)
## # A tibble: 3 x 4
##
     species
               average median standard_deviation
##
     <fct>
                        <dbl>
                 <dbl>
                                             <dbl>
## 1 Adelie
                  188.
                           188
                                             5.60
                           192
                                             5.75
## 2 Chinstrap
                  192.
## 3 Gentoo
                  213.
                           212
                                             3.90
ggplot(penguins.f,aes(penguins.f$flipper_length_mm,fill=species))+geom_boxplot()
```



ggplot(penguins.f,aes(penguins.f\$flipper_length_mm,fill=species))+geom_density()



Question 2:

- 1. Create a new data frame contains all the female with bill_depth_mm < 18. How many observations are
 - included in the new data frame.
- 2. Sort the data frame according to the value of bill depth.
- 3. Calculate the quantiles of the bill depth.
- 4. Define an indicator variable that takes the value of 1 for 16 < bill depth < 18 and zero otherwise. Produce a
 - 2×3 contingency table (see below) of the indicator and species. Use a chi-square test to test the hypothesis that the indicator and the species are independent.
- 5. Create a new data frame contains observations of female with bill depth between 16 to 18 mm. How many
 - observations are included in the new data frame.

```
library(palmerpenguins)
data("penguins", package = "palmerpenguins")
penguins <- drop_na(penguins)
penguins.f1<-penguins %>% filter(sex=="female")%>% filter(bill_depth_mm < 18)
dim(penguins.f1)</pre>
```

```
##
## index Adelie Chinstrap Gentoo
              26
                                58
##
       0
                          9
                         25
##
```

chisq.test(index,penguins.f1\$species)

```
##
##
   Pearson's Chi-squared test
##
## data: index and penguins.f1$species
## X-squared = 70.032, df = 2, p-value = 6.204e-16
```

Question 3:

In this question we focus on the variable bill_length_mm for the species Adelie and Chinstrap.

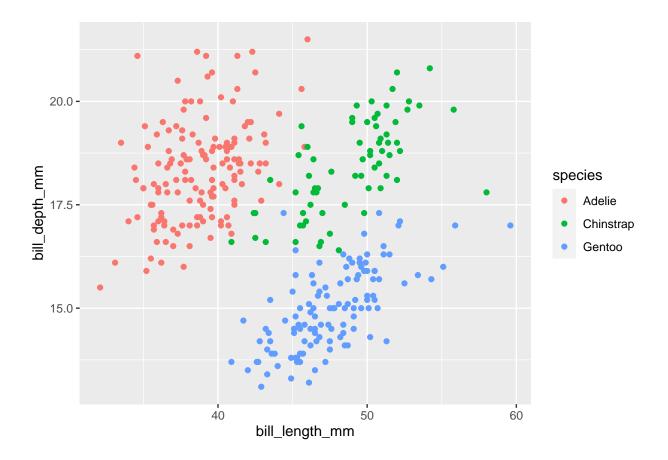
1. Produce the figures below and fit a regression model based on your figure.

Solution

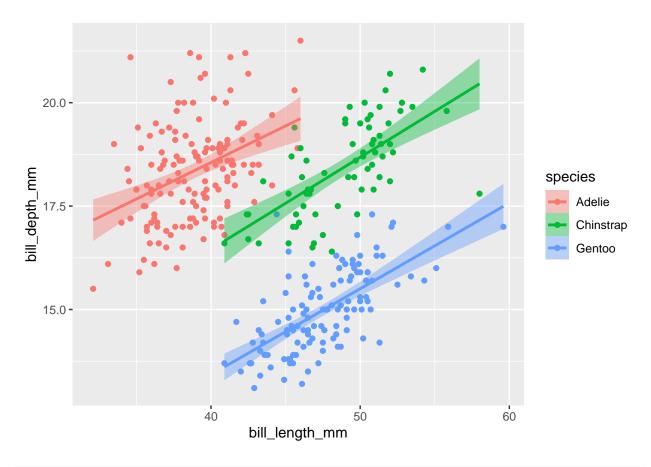
```
library(palmerpenguins)
data("penguins", package = "palmerpenguins")
penguins <- drop_na(penguins)
penguins.f<-penguins %>% filter(sex=="female")
penguins.fac<-penguins.f %>% filter(species%in%c("Adelie","Chinstrap"))
dim(penguins.fac)
```

[1] 107 8

```
#print(penguins.fac$species)
#tapply(penguins.fac$bill_length_mm,penguins.fac$species,mean)
ggplot(penguins,aes(bill_length_mm,bill_depth_mm,fill=species,color=species))+geom_point()
```



ggplot(penguins,aes(bill_length_mm,bill_depth_mm,fill=species,color=species))+geom_point()+geom_smooth(species)



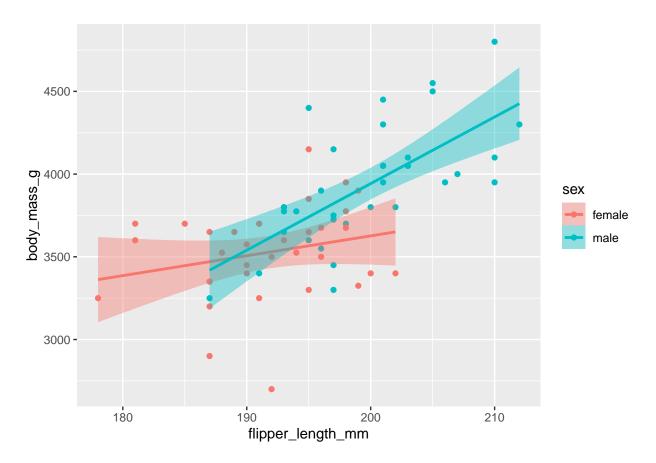
fit.1<-lm(bill_depth_mm~bill_length_mm+species,data=penguins)
summary(fit.1)</pre>

```
##
## Call:
## lm(formula = bill_depth_mm ~ bill_length_mm + species, data = penguins)
##
## Residuals:
##
      Min
               1Q Median
                               ЗQ
                                      Max
## -2.4579 -0.6814 -0.0431 0.5441 3.5994
##
## Coefficients:
##
                   Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                               0.69093 15.291 < 2e-16 ***
                   10.56526
                               0.01768 11.337 < 2e-16 ***
## bill_length_mm
                    0.20044
## speciesChinstrap -1.93308
                               0.22572 -8.564 4.26e-16 ***
                               0.19440 -26.252 < 2e-16 ***
## speciesGentoo
                   -5.10332
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Residual standard error: 0.9543 on 329 degrees of freedom
## Multiple R-squared: 0.7673, Adjusted R-squared: 0.7652
## F-statistic: 361.6 on 3 and 329 DF, p-value: < 2.2e-16
```

Question 4:

- 1. Create a new data frame contains only the Chinstrap species.
- 2. Fit a linear regression model in which the body mass is the response and flipper length and gender are the predictors.
- 3. Visualize your model.

```
library(palmerpenguins)
data("penguins", package = "palmerpenguins")
penguins <- drop_na(penguins)
penguins.c<-penguins %>% filter(species%in%c("Chinstrap"))
ggplot(penguins.c,aes(flipper_length_mm,body_mass_g,fill=sex,color=sex))+geom_point()+geom_smooth(method)
```



```
fit.2<-lm(body_mass_g~sex+flipper_length_mm:sex,data=penguins.c)
summary(fit.2)</pre>
```

```
##
## Call:
## lm(formula = body_mass_g ~ sex + flipper_length_mm:sex, data = penguins.c)
##
## Residuals:
```

```
##
               1Q Median
                                3Q
## -830.38 -171.79 -13.33 174.34 658.82
##
## Coefficients:
##
                               Estimate Std. Error t value Pr(>|t|)
                                                      0.759
                                                              0.4504
## (Intercept)
                               1224.901
                                           1613.134
                                           2285.691 -2.335
                                                              0.0227 *
## sexmale
                               -5336.249
## sexfemale:flipper_length_mm
                                  12.008
                                              8.410
                                                      1.428
                                                              0.1582
## sexmale:flipper_length_mm
                                  40.269
                                              8.097
                                                      4.974 5.22e-06 ***
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 278 on 64 degrees of freedom
## Multiple R-squared: 0.5003, Adjusted R-squared: 0.4769
## F-statistic: 21.36 on 3 and 64 DF, p-value: 1.059e-09
```

Question 5:

- 1. Create a new data frame contains all observations of the Biscoe island.
- 2. Sort the new data by body mass g.

188.7955

- 3. Sort the new data by gender and body_mass_g.
- 4. For the new data:
- (a) Test that the mean flipper length is equal cross the species levels.
- (b) Produce the figure below.
- (c) Use the chi-square test to test the hypothesis that species and gender are independent and produce the table below.

Solution

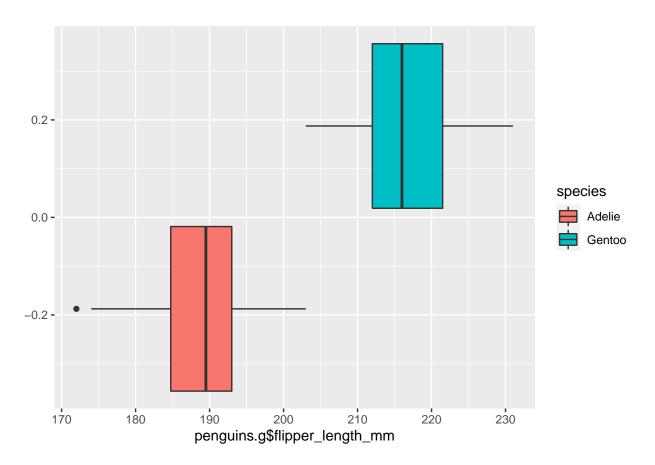
##

```
library(palmerpenguins)
data("penguins", package = "palmerpenguins")
penguins <- drop_na(penguins)
#head(penguins)
penguins.g<- penguins %>% filter(island%in%c("Biscoe"))
#penguins.g$species
#penguins.g$island
#quantile(penguins.g)
t.test(penguins.g$flipper_length_mm~penguins.g$species)
```

```
##
## Welch Two Sample t-test
##
## data: penguins.g$flipper_length_mm by penguins.g$species
## t = -24.091, df = 75.4, p-value < 2.2e-16
## alternative hypothesis: true difference in means between group Adelie and group Gentoo is not equal
## 95 percent confidence interval:
## -30.79132 -26.08836
## sample estimates:
## mean in group Adelie mean in group Gentoo</pre>
```

217.2353

```
ggplot(penguins.g,aes(penguins.g$flipper_length_mm,fill=species))+geom_boxplot()
```



```
xx<-table(penguins.g$sex,penguins.g$species)
xx1<-xx[,-c(2)]
xx1</pre>
```

```
## ## Adelie Gentoo
## female 22 58
## male 22 61
```

chisq.test(penguins.g\$sex,penguins.g\$species)

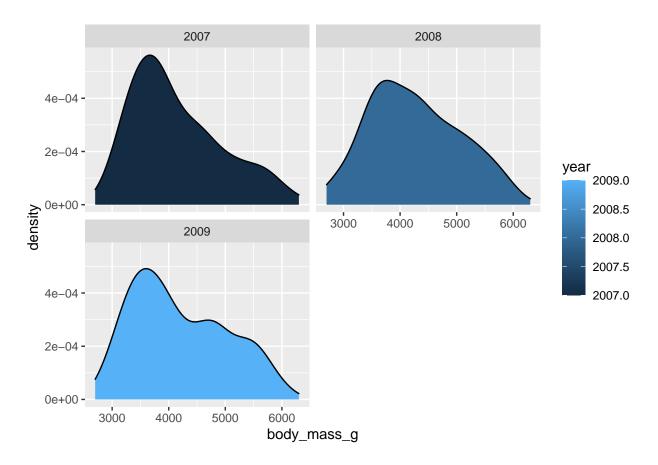
```
##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data: penguins.g$sex and penguins.g$species
## X-squared = 1.1478e-30, df = 1, p-value = 1
```

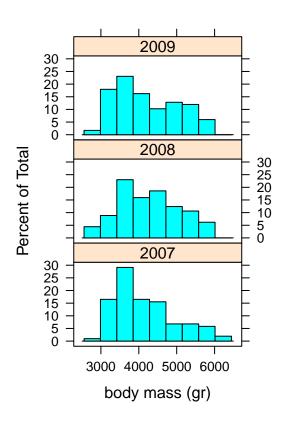
Question 6:

- 1. In the penguins data, how many observations there are at each year.
- 2. Produce the multi-way density and histogram below.

- 3. Create a new dataset with observations before 2009 and include in the data the year, sex, body_mass_g.
- 4. Define a new indicator variable which take the value of 1 for observations with body mass > 4000 and zero otherwise.
- 5. Produce a 2×2 table (year X indicator) for the overall sample, for male and for female (separately).

```
library(palmerpenguins)
data("penguins", package = "palmerpenguins")
penguins <- drop_na(penguins)
ggplot(penguins, aes(body_mass_g,fill = year)) +
geom_density() +
facet_wrap(~year,ncol = 2)</pre>
```





```
#table(penguins$year)
#quantile(penguins$body_mass_g)
penguins.g<- penguins %>% filter(year < 2009)</pre>
body_m<-penguins.g$body_mass_g*0+1
body_m[penguins.g$body_mass_g <4001]<-0
table(body_m,penguins.g$year) # all sample
##
## body_m 2007 2008
##
            56
                  49
##
        1
            47
                  64
penguins.gf<- penguins.g %>% filter(sex=="female")
#dim(penguins.gf)
body_mf<-body_m[penguins.g$sex=="female"]</pre>
#length(body_mf)
table(body_mf,penguins.gf$year) #female
##
## body_mf 2007 2008
##
         0
             34
                   35
##
         1
             17
                   21
```

```
penguins.gm<- penguins.g %>% filter(sex=="male")
#dim(penguins.gm)
body_mm<-body_m[penguins.g$sex=="male"]
#length(body_mm)
table(body_mm,penguins.gm$year) #male

##
## body_mm 2007 2008
## 0 22 14</pre>
```

1 30

43

##