

Statistics for Computer Science

Assignment 1

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Exercise 1

First, I loaded the data. Then I made a subset of cranial breadth of males from populations AUSTRALI and PERU. This was unsorted, so I sorted it and checked for missing values. There were no missing values.

```
1 # I commented setwd because you probably don't have this directory on
  your machine.
2 # And .Rnw needs to be compilable on your machine.
3 #setwd("/home/martingregorik/R/assignment01")
4 options(max.print=10000)
5 library(xtable)
6
7 howell <- read.csv("Howell.csv", header = TRUE)
8 #str(howell)
9
10 xcb_unsort <- howell$XCB[howell$Sex == 'M' & (howell$Population == '
    AUSTRALI' | howell$Population == 'PERU')]
11 xcb <- sort(xcb_unsort)
12 #sum(is.na(xcb)) = 0
13 #is.na(xcb)
```

Then I created custom functions.

```
14 # Task 1
15 my_sample_min <- function(vec) {
16   min <- Inf
17   for (i in 1:length(vec)) {
18     if (vec[i] < min) {
19       min <- vec[i]
20     }
21   }
22   return (min)
23 }
24
25 my_sample_max <- function(vec) {
26   max <- -Inf
27   for (i in 1:length(vec)) {
28     if (vec[i] > max) {
29       max <- vec[i]
30     }
31   }
32   return (max)
33 }
34
35 my_sample_mean <- function(vec) {
36   sum <- 0
37   for (i in 1:length(vec)) {
38     sum <- sum + vec[i]
39   }
```

```
40   return (sum / length(vec))
41 }
42
43 # sum of (xi - x~) squared
44 my_sum_sample_avg <- function(vec, exponent=2) {
45   sum <- 0
46   div <- 0
47   avg <- my_sample_mean(vec)
48   for (i in 1:length(vec)) {
49     div <- vec[i] - avg
50     sum <- sum + (div ^ exponent)
51   }
52   return (sum)
53 }
54
55 my_decile <- function(vec, k) {
56   # k / 10 * 100
57   return (vec[1:(k * 10)])
58 }
59
60 my_quartile <- function(vec, q) {
61   # denominator
62   denom <- 1 / q
63   len <- length(vec)
64   if (len %% 2 == 0) {
65     # even
66     return ((vec[len / denom] + vec[len / denom + 1]) / 2)
67   } else {
68     # odd
69     return (vec[(len + 1) / denom])
70   }
71 }
72 median <- my_quartile(xcb, 0.5)
73
74 my_five_num_sum <- function(vec) {
75   return (data.frame(min=my_sample_min(vec), lower_q=my_quartile(vec,
76     0.25), median=my_quartile(vec, 0.50), upper_q=my_quartile(vec,
77     0.75), max=my_sample_max(vec)))
78 }
79
80 my_sample_skew_cramer <- function(vec) {
81   nom <- my_sum_sample_avg(vec, 3)
82   denom <- length(vec) * (my_sample_variance(vec) ^ (3 / 2))
83   return (nom / denom)
84 }
85
86 my_sample_kurtosis <- function(vec) {
87   nom <- my_sum_sample_avg(vec, 4)
88   denom <- length(vec) * (my_sample_variance(vec) ^ 2)
```

```

87   return ((nom / denom) - 3)
88 }
89 # broad = thick tails = platykurtic
90
91 my_sample_variance <- function(vec, exponent=2) {
92   if (length(vec) != 0) {
93     avg <- my_sample_mean(vec)
94     sum <- my_sum_sample_avg(vec, exponent)
95     return (sum / (length(vec) - 1))
96   }
97 }
98
99 my_sample_standard_deviation <- function(vec, exponent=2) {
100   if (length(vec) != 0) {
101     pw <- my_sample_variance(vec, exponent)
102     return (sqrt(pw))
103   }
104 }
105
106 my_sample_range <- function(vec) {
107   return (my_sample_max(vec) - my_sample_min(vec))
108 }
109
110 my_sample_decile_range <- function(vec) {
111   return (my_quartile(vec, 0.90) - my_quartile(vec, 0.10))
112 }
113
114 my_sample_trimmed_avg <- function(vec) {
115   gamma <- 0.1
116   n <- length(vec)
117   g <- floor(gamma * n)
118   # xtg
119   return ((1 / (n - 2 * g)) * sum(vec[g + 1:n - g]))
120 }
121
122 my_sample_trimmed_var <- function(vec) {
123   gamma <- 0.1
124   n <- length(vec)
125   g <- floor(gamma * n)
126   xtg <- my_sample_trimmed_avg(vec)
127   # stg
128   return ((1 / (n - (2 * g) - 1)) * sum((vec[g + 1:n - g] - xtg)^2))
129 }

```

Then I calculated characteristics of each population and stored them in a table.

```

130 # Australia
131 xcb_australi_unsorted <- howell$XCB[howell$Sex == 'M' & howell$
    Population == 'AUSTRALI']
132 xcb_australi <- sort(xcb_australi_unsorted)

```

```
133 xcb_aus_tab <- round(data.frame(
134   size=length(xcb_australi),
135   mean=my_sample_mean(xcb_australi),
136   my_five_num_sum(xcb_australi),
137   skew=my_sample_skew_cramer(xcb_australi),
138   kurt=my_sample_kurtosis(xcb_australi),
139   variance=my_sample_variance(xcb_australi),
140   sd=my_sample_standard_deviation(xcb_australi),
141   range=my_sample_range(xcb_australi),
142   dec_range=my_sample_decile_range(xcb_australi),
143   trim_avg=my_sample_trimmed_avg(xcb_australi),
144   trim_var=my_sample_trimmed_var(xcb_australi)
145 ), 4)
146 # Peru
147 xcb_peru_unsorted <- howell$XCB[howell$Sex == 'M' & howell$Population ==
    'PERU']
148 xcb_peru <- sort(xcb_peru_unsorted)
149 xcb_peru_tab <- round(data.frame(
150   size=length(xcb_peru),
151   mean=my_sample_mean(xcb_peru),
152   my_five_num_sum(xcb_peru),
153   skew=my_sample_skew_cramer(xcb_peru),
154   kurt=my_sample_kurtosis(xcb_peru),
155   variance=my_sample_variance(xcb_peru),
156   sd=my_sample_standard_deviation(xcb_peru),
157   range=my_sample_range(xcb_peru),
158   dec_range=my_sample_decile_range(xcb_peru),
159   trim_avg=my_sample_trimmed_avg(xcb_peru),
160   trim_var=my_sample_trimmed_var(xcb_peru)
161 ), 4)
162 # Concat them to single data frame
163 xcb_tab <- xcb_aus_tab
164 xcb_tab[2, ] <- xcb_peru_tab
165 rownames(xcb_tab) <- c("AUSTRALI", "PERU")
166 # Since the table is too long for pdf, I split them into two.
167 # Yes, I concatenated them earlier, but that was rows.
168 # Now I cut them in half by columns.
169 xcb_tab_first_half <- xcb_tab[, 1:(length(xcb_tab) / 2)]
170 xcb_tab_second_half <- xcb_tab[, ((length(xcb_tab) / 2) + 1):length(xcb_
    tab)]
```

	size	mean	min	lower _q	median	upper _q	max
AUSTRALI	52.00	131.94	124.00	128.00	131.00	134.00	144.00
PERU	55.00	137.95	129.00	135.00	138.00	141.00	149.00

Table 1: Characteristics of maximal cranial breadth of AUSTRALI and PERU populations

	skew	kurt	variance	sd	range	dec _r ange	trim _a vg
AUSTRALI	0.64	-0.34	26.06	5.10	20.00	14.00	163.36
PERU	-0.00	0.06	15.87	3.98	20.00	9.00	168.60

Table 2: Characteristics of maximal cranial breadth of AUSTRALI and PERU populations

Next, I created boxplots of maximum cranial breadth for each population. I set the width of boxes to be proportional to sample sizes.

```

171 # Different lengths (australi is 52, peru 55), need to add 0s to the end(
    neglected in output)
172 n <- max(length(xcb_australi), length(xcb_peru))
173 xcb_australi_prolonged <- xcb_australi
174 length(xcb_australi_prolonged) <- n
175 max_b <- data.frame(AUSTRALI=xcb_australi_prolonged, PERU=xcb_peru)

176 # Variable australi_peru_cols will be used in following exercises
177 australi_peru_cols = c("dodgerblue4", "indianred")
178 boxplot(
179   max_b,
180   width = c(length(xcb_australi), length(xcb_peru)),
181   notch = TRUE,
182   main = "Boxplot of maximal cranial breadth",
183   xlab = "Countries",
184   ylab = "Maximal cranial breadth (mm)",
185   col = australi_peru_cols,
186   pch = 16
187 )
188 points(1:2, xcb_tab$mean, col = "green", pch = 16)
189 legend('topleft', pch = 16, legend = c("Maximum", "Average"), col = c("
    black", "green"))

```

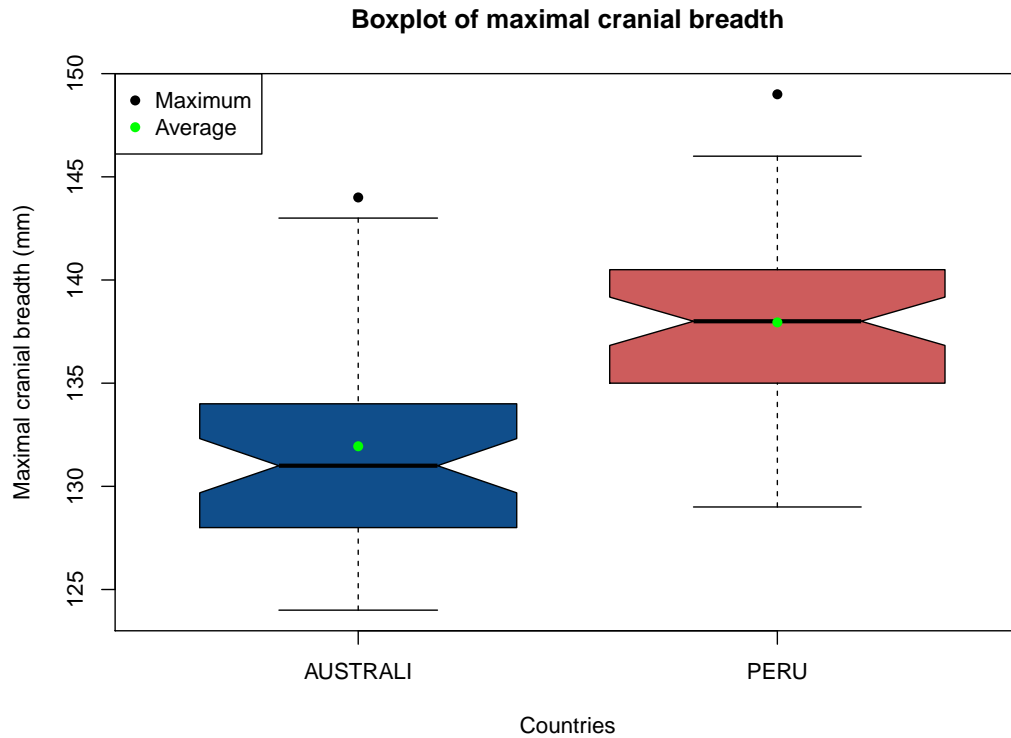


Figure 1: Boxplot of maximal cranial breadth(mm)

Created histogram of maximum cranial breadth for each population.

```

190 def.par <- par()
191 layout(matrix(c(1, 2), nrow=1, ncol=2), widths=c(0.5, 0.5))
192 #layout.show(n=2)
193 hist(
194   xcb_australi,
195   main = "Australi population",
196   xlab = "Maximal cranial breadth(mm)",
197   ylab = "Count",
198   col = australi_peru_cols[1]
199 )
200 hist(
201   xcb_peru,
202   main = "Peru population",
203   xlab = "Maximal cranial breadth(mm)",
204   ylab = "Count",
205   col = australi_peru_cols[2]
206 )

```

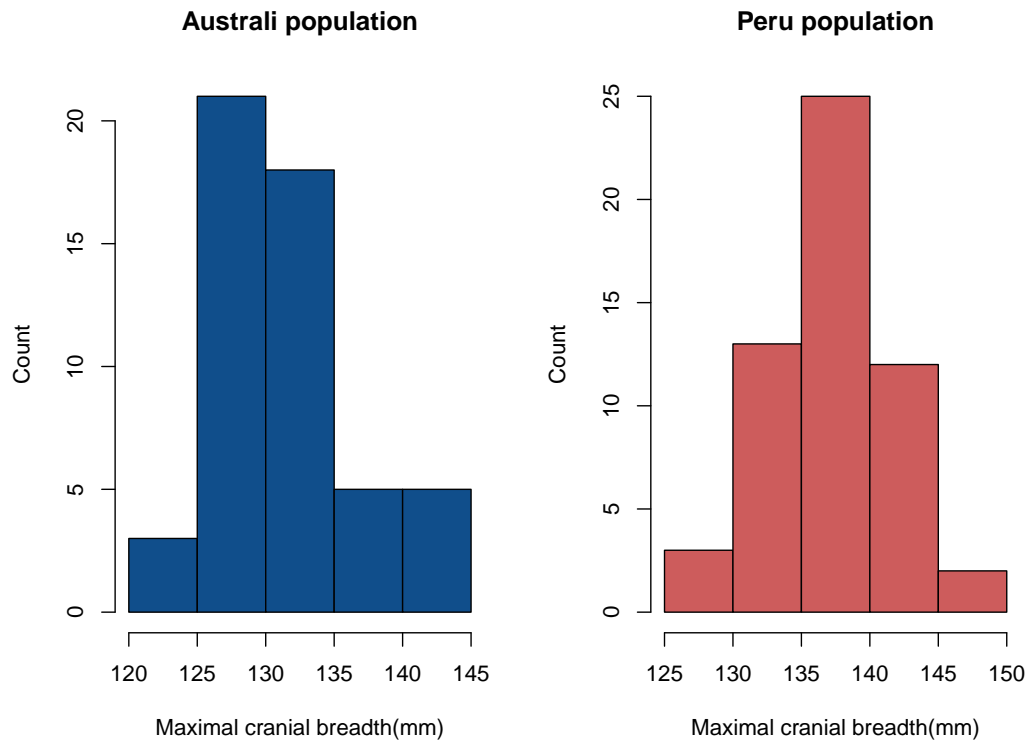


Figure 2: Histogram of maximal cranial breadth(mm)

Created normal qq-plot of maximum cranial breadth for each population.

The observed measurements(Y axis values) are from the same range so the plots can be easily compared.

```

207 layout(matrix(c(1, 2), nrow=1, ncol=2), widths=c(0.5, 0.5))
208 #layout.show(n=2)
209 qqnorm(
210   y = xcb_australi,
211   main = "Australi population",
212   xlab = "Theoretical quantities",
213   ylab = "Maximal cranial breadth(mm)",
214   ylim = c(min(xcb_australi, xcb_peru), max(xcb_australi, xcb_peru)),
215   pch = 16,
216   col = australi_peru_cols[1]
217 )
218 qqline(xcb_australi)
219 qqnorm(
220   y = xcb_peru,
221   main = "Peru population",
222   xlab = "Theoretical quantities",
223   ylab = "Maximal cranial breadth(mm)",
224   ylim = c(min(xcb_australi, xcb_peru), max(xcb_australi, xcb_peru)),
225   pch = 16,
226   col = australi_peru_cols[2]

```



```
227 )  
228 qqline(xcb_peru)
```

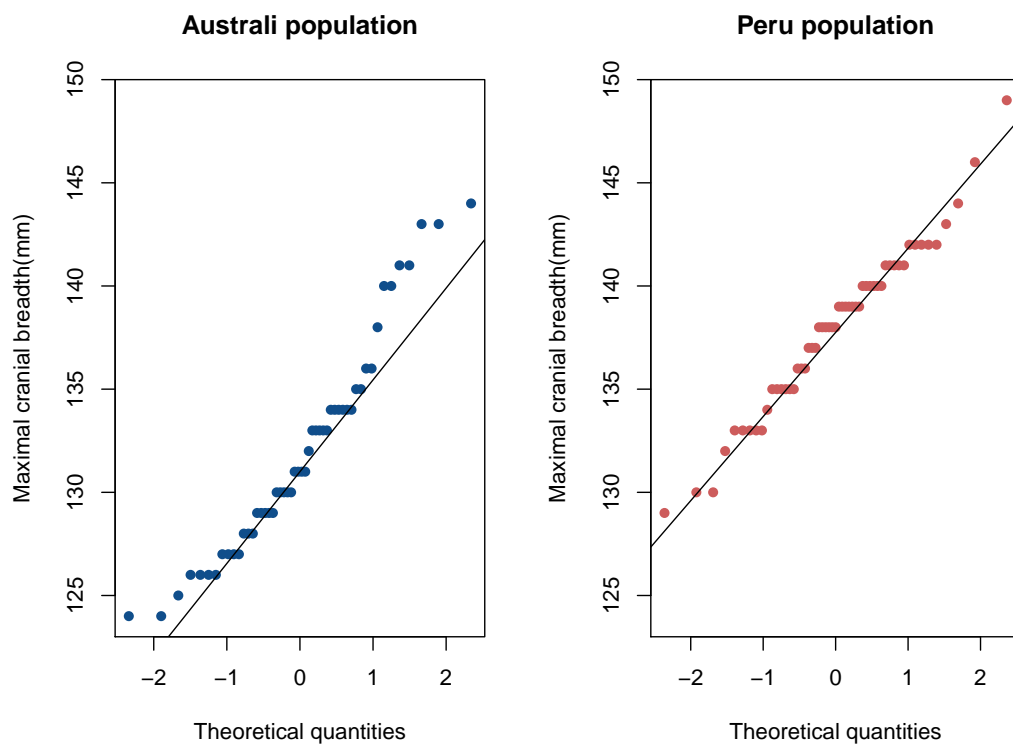


Figure 3: Normal qq-plot of maximal cranial breadth(mm)

Results and interpretation

Although both populations have same maximal cranial breadth reached, the differences can be best seen in histograms - more people of Peru population have bigger breadth than Australian people.

Variable XCB, representing maximal cranial breadth of each population, seems to be normally distributed variable in both cases.

Exercise 2

First, I loaded the data. Then I checked for missing values in desired subset. There were none. So I summed the values in each column, which gave me a total sum of men and women in each year. Lastly, I added margins, which gave me total population in all of the years. Also, I like to keep variables at the beginning of each significant part, in this case Exercise.

```

229 area_esp <- read.csv("area_spanish_provinces.csv", header = TRUE)
230 pop_esp <- read.csv2("population-spain-1998-2018.csv", header = TRUE)
231 #str(area_esp)
232 #str(pop_esp)
233 #pop_esp
234 #sum(is.na(pop_esp[, 2:length(pop_esp)])) = 0
235 ### VARIABLES ###
236 legend_sex <- c("Women", "Men")
237 stat_years <- c("1998", "2018")
238 total_pop_cols = c("pink", "dodgerblue")
239 year_cols = c("darkgoldenrod", "chocolate")
240 ### VARIABLES ###
241 people_each_year <- as.table(colSums(pop_esp[, 2:length(pop_esp)]))
242 total_people_each_year <- addmargins(people_each_year, FUN = c(Total=sum
  ))

```

Then I created a barplot of total population of Spain in each of the years, with each bar divided between men and women.

The data were stored as table(from previous task), so I converted them into data frame to be able to subset values.

The years were descending from 2018, so I made them ascending.

Then I stored the data in matrix in order for barplot to accept the argument type. Since the data in data frame is ordered by gender, I simply added "byrow = TRUE" to the matrix in order to store the data correctly.

I have also set a few plot parameters:

"cex.axis", to make Y axis values smaller,

"las", to make axis labels horizontal,

"bty" in legend to hide legend box.

```

243 df_people <- as.data.frame(people_each_year)
244 # switched ordering of years
245 df_people <- df_people[dim(df_people)[1]:1, ]
246 df_people <- matrix(
247   df_people[, 2],
248   nrow = 2,
249   ncol = 5,
250   byrow = TRUE,
251   dimnames = list(c("F", "M"), c("1998", "2003", "2008", "2013", "2018"))
252 )
253 # It is clean now
254 suppressWarnings(par(def.par))
255 barplot(

```

```
256 height = df_people,  
257 main = "Total population of Spain",  
258 ylim = c(0, max(df_people)),  
259 beside = TRUE,  
260 las = 1,  
261 cex.axis = 0.8,  
262 col = total_pop_cols)  
263 legend('topleft', pch = 15, legend = legend_sex, col = total_pop_cols,  
      bty = 'n')
```

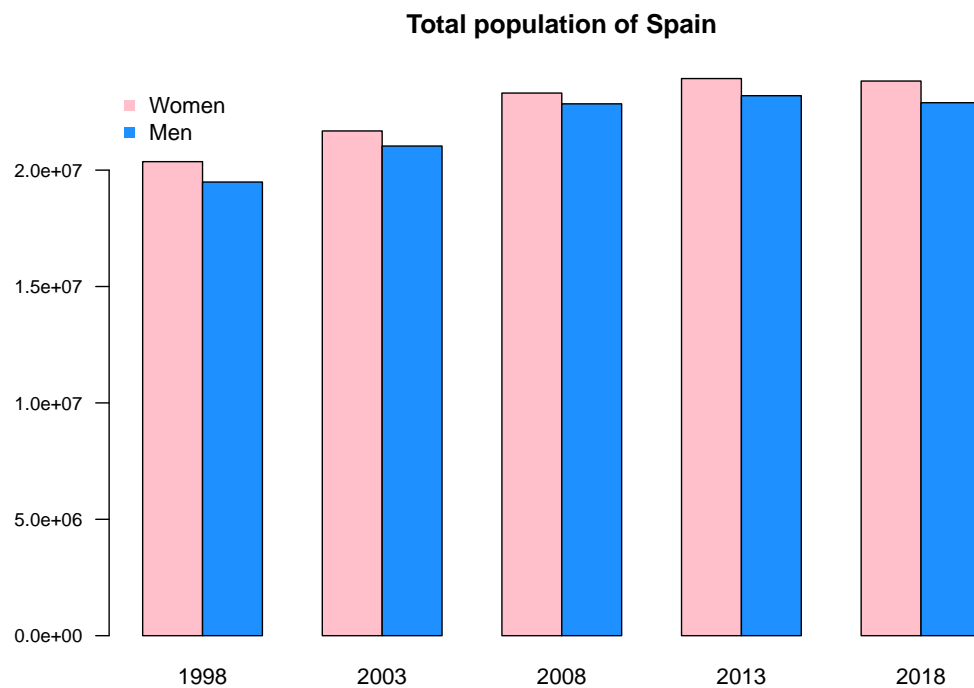


Figure 4: Barplot of total population in Spain

Then I created a barplot of relative proportions of men and women in each province in 2018. I transposed the input matrix in order to show the data by row (men and women side-by-side). I have also set few plot parameters:

"cex.names", to make the provinces' names smaller,

"las", to make axis labels perpendicular to the axis.

```

264 relative_pop_2018_provinces <- matrix(
265   data = c(pop_esp$females.2018, pop_esp$males.2018),
266   nrow = 52,
267   ncol = 2
268 )
269 barplot(
270   height = t(relative_pop_2018_provinces),
271   names.arg = pop_esp$province,
272   cex.names = 0.6,
273   cex.axis = 0.8,
274   main = "2018 relative population of Spain",
275   ylim = c(0, max(relative_pop_2018_provinces)),
276   las = 2,
277   beside = TRUE,
278   col = total_pop_cols)
279 legend('topleft', pch = 15, legend = legend_sex, col = total_pop_cols,
      bty = 'n')

```

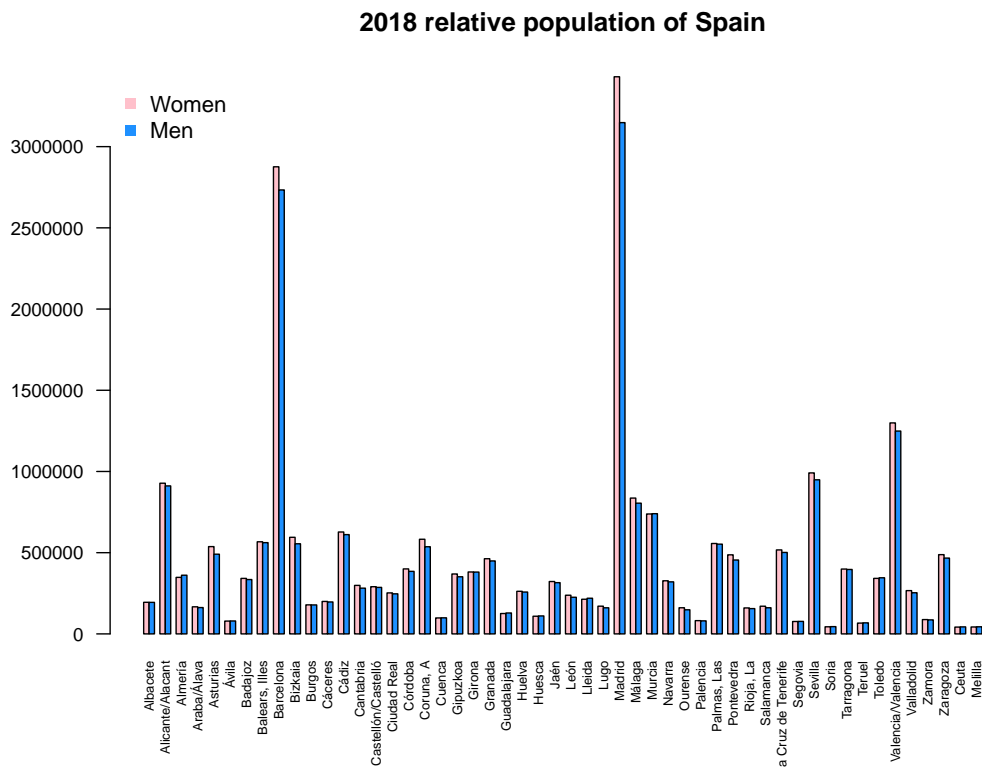


Figure 5: Barplot of relative population of Spain in 2018

Then I calculated population density in 1998 and 2018, for each province. The area of each province was loaded in the beginning of this exercise. I created a helper matrix to store intermediate values.

```

280 ppl_in_two_years <- matrix(
281   data = c(
282     pop_esp$males.1998,
283     pop_esp$females.1998,
284     pop_esp$males.2018,
285     pop_esp$females.2018
286   ),
287   nrow = nrow(pop_esp),
288   ncol = 4,
289   dimnames = list(pop_esp$province, c("males.1998", "females.1998", "
      males.2018", "females.2018")))
290 )
291 total_1998 <- as.vector(rowSums(ppl_in_two_years[, 1:2]))
292 total_2018 <- as.vector(rowSums(ppl_in_two_years[, 3:4]))
293 density_1998 <- total_1998 / area_esp$Area
294 density_2018 <- total_2018 / area_esp$Area
295 den <- data.frame(
296   Province=pop_esp$province,
297   "Year 1998" = density_1998,
298   "Year 2018" = density_2018
299 )

```

In subtask a, firstly, I sorted the data by year. Next, I calculated the estimates of characteristics of population density, each year separately. Lastly, I bound them together into one table.

```

300 # subtask a)
301 den_sorted_1998 <- sort(den$Year.1998)
302 den_sorted_1998_tab <- round(data.frame(
303   size=length(den_sorted_1998),
304   mean=my_sample_mean(den_sorted_1998),
305   my_five_num_sum(den_sorted_1998),
306   skew=my_sample_skew_cramer(den_sorted_1998),
307   kurt=my_sample_kurtosis(den_sorted_1998),
308   sd=my_sample_standard_deviation(den_sorted_1998)
309 ), 4)
310 den_sorted_2018 <- sort(den$Year.2018)
311 den_sorted_2018_tab <- round(data.frame(
312   size=length(den_sorted_2018),
313   mean=my_sample_mean(den_sorted_2018),
314   my_five_num_sum(den_sorted_2018),
315   skew=my_sample_skew_cramer(den_sorted_2018),
316   kurt=my_sample_kurtosis(den_sorted_2018),
317   sd=my_sample_standard_deviation(den_sorted_2018)
318 ), 4)
319 density_characteristics <- rbind(den_sorted_1998_tab, den_sorted_2018_
      tab)
320 row.names(density_characteristics) <- stat_years

```

	size	mean	min	lower _q	median	upper _q	max	skew	kurt	sd
1998	52.00	282.01	9.17	29.36	79.29	118.94	4623.69	4.12	16.24	848.58
2018	52.00	356.87	9.02	28.23	84.74	151.19	6644.92	4.38	19.20	1128.98

Table 3: Characteristics of population density of Spain in 1998 and 2018

In subtask b, I created a boxplot of population density in 1998 and in 2018. I have set the Y axis limits in order to see the graphs nicely, because there were some outliers which would squish the whole plot.

```

321 # subtask b)
322 suppressWarnings(par(def.par))
323 boxplot(
324   den[2:3],
325   notch = TRUE,
326   main = "Pop. density in 1998 & 2018",
327   ylab = "Population density",
328   ylim = c(0, 250),
329   pch = 16,
330   col = year_cols
331 )

```

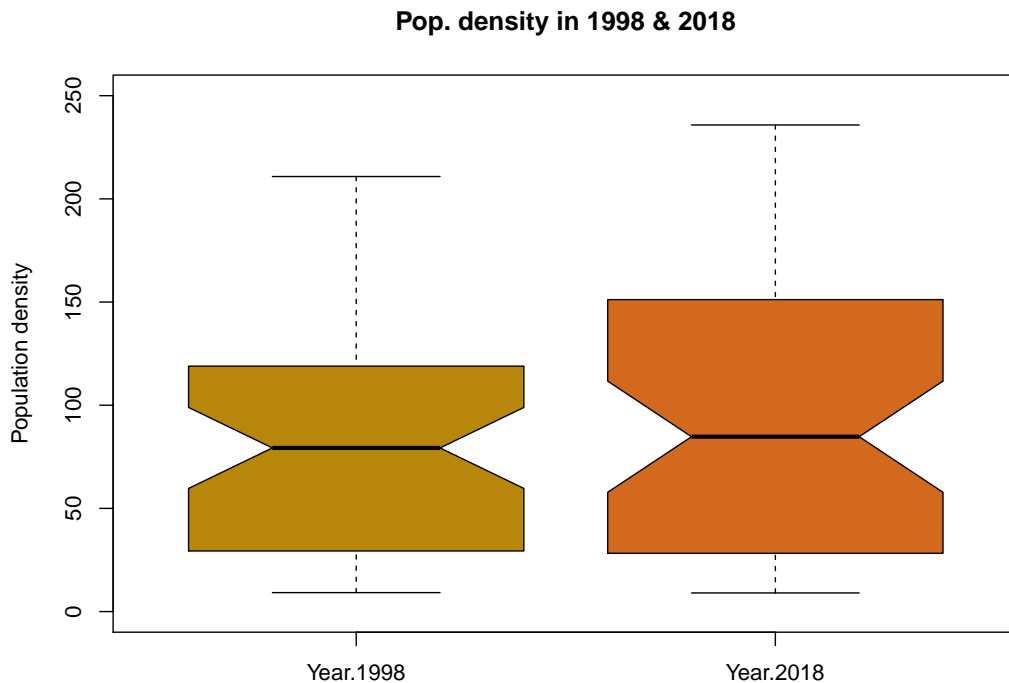


Figure 6: Boxplot of population density in Spain in 1998 and 2018

In subtask c, I created histogram of population density in 1998 and in 2018. There are two histograms, one for each year. Histogram for year 2018 is overlapping histogram for year 1998.

I changed the number of breaks. I was hoping to achieve a plot with province names, just like barplot of relative population of Spain(Figure 5), but it did not turn out that way. I'm really sad because that would make comparisons far better. Now I can't recognize the provinces. I left it as it is because I still think it looks better than the default.

Lastly, I also changed axes a little bit because I didn't like the values inserted by default, they were too sparse, too spread out.

```

332 # subtask c)
333 suppressWarnings(par(def.par))
334 rgb_cols = c(rgb(0.6,0.4,0.2,0.5), rgb(0.3,0.5,0.6,0.5))
335 hist(
336   den$Year.1998,
337   main = "Pop. density in 1998 & 2018",
338   xlab = "Density(people/km^2)",
339   ylab = "Number of provinces",
340   las = 2,
341   col = rgb_cols[1],
342   breaks = density_characteristics$size[1],
343   xaxt = 'n',
344   yaxt = 'n'
345 )
346 # X axis
347 axis(
348   side = 1,
349   at = seq(0, max(density_characteristics$max), by = 250),
350   las = 2,
351   cex.axis = 0.8
352 )
353 # Y axis
354 axis(
355   side = 2,
356   at = seq(0, density_characteristics$size[1], by = 5),
357   las = 2,
358   cex.axis = 0.8
359 )
360 hist(
361   den$Year.2018,
362   col = rgb_cols[2],
363   breaks = density_characteristics$size[2],
364   add = T)
365 legend('topright', pch = 15, legend = stat_years, col = rgb_cols, bty =
    'n')

```

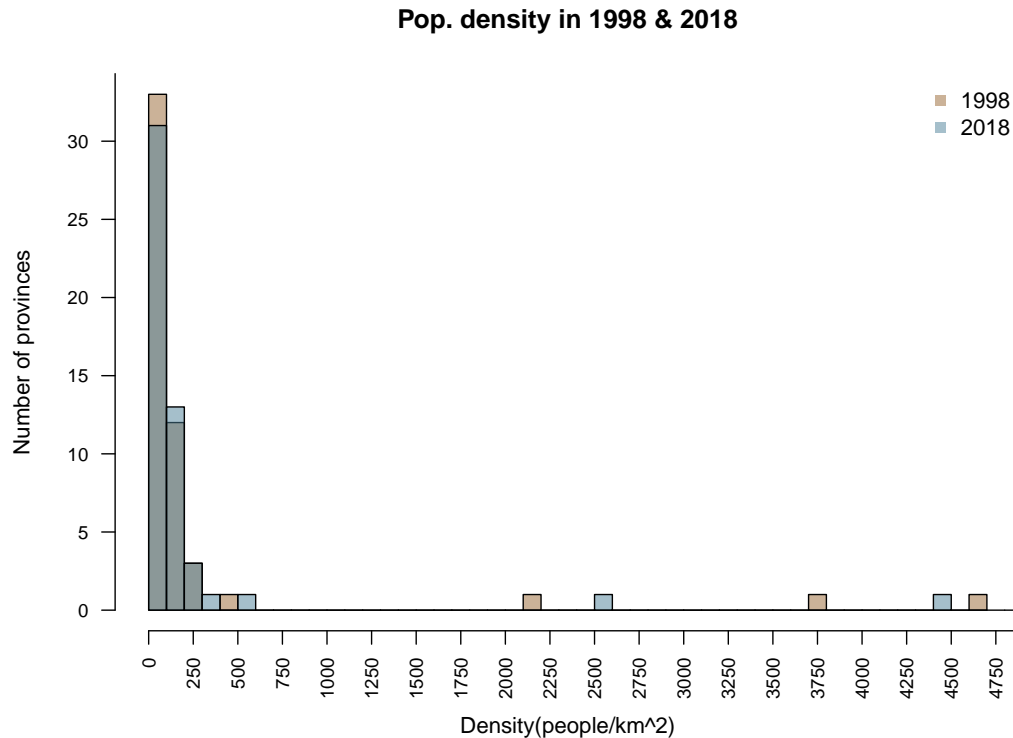



Figure 7: Histogram of population density in Spain in 1998 and 2018

Results and interpretation

Total population of Spain increased in years 1998-2018, even though there is a minor decrease between years 2013 and 2018.

Women are still slightly dominant when it comes to population proportions.

As we can see from the population density characteristics, the maximum population density increased from 4600 to 6600, which can also be seen in histogram. The mean population density increased from 282 to 356, which can be seen better e.g. in boxplot.