# European Doctoral School of Demography (EDSD) Computer Programming E140

Özer Bakar Liliana Patricia Calderon Bernal Gonzalo Daniel Garcia Ainhoa-Elena Leger Özge Elif Özer

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#### Exercise 1

The first step to work with the practice data set of the German Socio Economic Panel (SOEP) is to download and unzip it.

```
# Download and unzip the file
soep_url <-
   "https://www.diw.de/documents/dokumentenarchiv/17/diw_01.c.412698.de/soep_lebensz_en.zip"
destfile <- "soep_lebensz_en.zip"
download.file(soep_url, destfile)
unzip(zipfile = "soep_lebensz_en.zip")</pre>
```

#### 1a) Load the data set into R

We load the foreign and tidyverse packages which will be useful to solve the exercises. The read.dta() function allows us to load the Stata's database into R.

```
# Loading necessary libraries to solve the assignment
#install.packages("foreign")
#install.packages("tidyverse")

library(foreign)
library(tidyverse)

# Importing the Stata data into R framework by using foreign library
soep <- read.dta("soep_lebensz_en.dta", convert.factors = TRUE)</pre>
```

Let's have a look whether the data set has been correctly imported.

```
glimpse(soep)
```

The practice data set consists of a total of 9 variables and 12922 observations. Two variables identify the individuals and the year of the survey, while the other collect information about sex, education measured in years, number of kids, perceived subjective health, satisfaction in life and their standardized versions.

## 1b) How many unique individuals are included in the practice data set?

We just need to find out how many identification number ID are in the data set, as they are unique for each individual. The function unique() returns the vector with duplicate elements removed and the function length() returns the length of that vector.

```
soep$id %>% unique %>% length

## [1] 3550
The same result can be obtained with base R.
length(unique(soep$id))
```

There are 3550 distinct individuals in the data set.

## 1c) Tabulate the number of observations per year

We group the observations by year and count the observation for each year thanks to the function tally().

```
## # A tibble: 5 x 2
##
      year
                n
##
     <int> <int>
## 1
      2000
            3198
## 2
      2001
            2690
## 3
      2002
            2485
## 4
      2003
             2299
## 5
      2004
            2250
```

There are 5 years of measurement and more observations in the most distant years. An alternative in base R is to use the function table(). To have an idea of the relative frequencies, we apply the function prop.table().

```
prop.table(table(soep$year))

##

## 2000 2001 2002 2003 2004

## 0.2474849 0.2081721 0.1923077 0.1779136 0.1741217
```

The observations in 2000 constitute the 25% of the data set and the ones in 2004 the 17% of the data set.

#### 1d) Restrict the data to the most recent year

Both the functions filter() and subset() return a subset of the data and can be used to retain all rows for which the year is the last available.

As expected, the new data set contains the same 9 variables but 2250 observations from 2004.

# What is the proportion of females in this subset of the data?

The function prop.table() used together with the function table() gives us the relative frequencies of females and males.

```
last_soep$sex %>% table %>% prop.table

## .
## male female
## 0.4577778 0.5422222
```

About 54.22% of the surveyed individuals are females.

# Is the average subjective health higher for men or for women?

To obtain the average subjective health for men or for women we first need to create a numerical variable from the categorical variable health\_org. We look at the levels of health\_org and create our numerical variable.

```
# Levels of the variable health org
levels(last_soep$health_org)
## [1] "not valid"
                           "does not concern" "no answer"
                                                                  "bad"
## [5] "poor"
                           "satisfactory"
                                                                  "very good"
# Creation of the corresponding numerical variable
last_soep$health_num[last_soep$health_org == c("not valid",
                                                "does not concern",
                                                "no answer")] <- 0
last_soep$health_num[last_soep$health_org =="bad"]
                                                              <- 1
last_soep$health_num[last_soep$health_org =="poor"]
                                                              <- 2
last_soep$health_num[last_soep$health_org =="satisfactory"] <- 3</pre>
last_soep$health_num[last_soep$health_org =="good"]
last soep$health num[last soep$health org =="very good"]
```

Let's double check whether the old variable and the new variable coincides.

```
table(last_soep$health_org, last_soep$health_num)
```

```
##
##
                                2
                                     3
                                         4
                                              5
                           1
##
     not valid
                           0
                                0
                                     0
                                         0
                                              0
##
                                0
                                     0
                                         0
                                              0
     does not concern
                           0
##
     no answer
                           0
                                0
                                     0
                                         0
                                             0
##
     bad
                          73
                                0
                                     0
                                         0
                                             0
##
                           0 306
                                     0
                                         0
                                             0
     poor
                                         0
##
     satisfactory
                           0
                                0 696
                                              0
##
                                0
                                    0 944
                                              0
     good
                           0
##
     very good
                                0
                                     0
                                         0 231
```

The creation of the new variable has worked. We can now compute the means for the females and the males. The function tapply() allows us to apply the mean to health\_num by group.

```
tapply(last_soep$health_num, last_soep$sex, mean)
```

```
## male female
## 3.464078 3.390164
```

The subjective health is perceived higher in men (3.46 on average) compared to women (3.39 on average).

The code below does the same operations and returns the same results.

## `summarise()` ungrouping output (override with `.groups` argument)

#### Exercise 2

### 2a) Load the data

We choose to analyze life expectancies at birth in Italy and we use the function readHMDweb from the package HMDHFDplus to read the data online from the The Human Mortality Database.

Please be aware that the user has to specify the personal credentials in the function.

```
#install.packages("HMDHFDplus")
library(HMDHFDplus)

# The user has to provide HMD username and password
italy_e0 <- readHMDweb("ITA","EOper","...")</pre>
```

The data loaded in R contains the life expectancies by gender and for the total population in Italy from 1872 to 2017. Below are the first and last rows of the data set.

```
head(italy_e0)
```

```
## Year Female Male Total
## 1 1872 30.26 29.28 29.76
## 2 1873 31.84 31.49 31.66
## 3 1874 32.02 31.62 31.81
## 4 1875 31.66 31.13 31.39
## 5 1876 34.00 33.36 33.67
## 6 1877 35.20 34.75 34.96
```

```
tail(italy_e0)
```

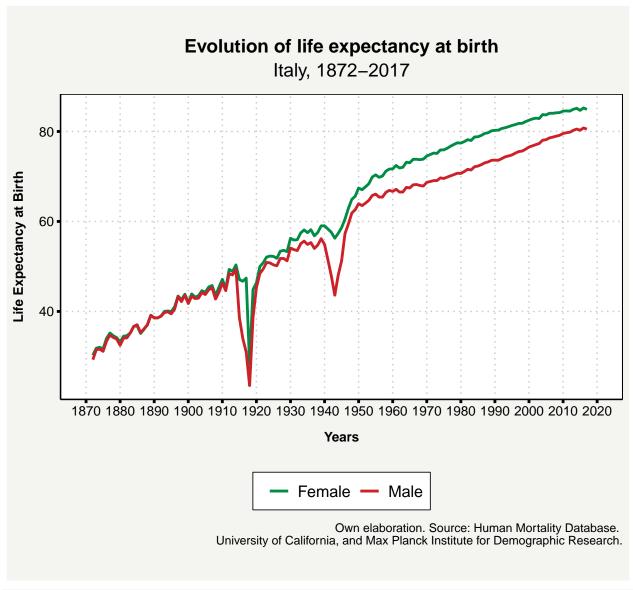
```
## Year Female Male Total
## 141 2012 84.53 79.83 82.28
## 142 2013 84.93 80.25 82.69
## 143 2014 85.15 80.54 82.95
## 144 2015 84.69 80.27 82.57
## 145 2016 85.23 80.79 83.10
## 146 2017 84.91 80.51 82.79
```

#### 2b) Visualize the trend in life expectancy at birth

In the following we plot the trends in life expectancy at birth for women and men.

```
# Customization of the theme
theme_graphs <- function (base_size = 16, base_family = "sans") {
  theme(plot.title = element_text(size = 14, face = "bold",
                                  hjust=0.5, margin = margin(20, 0, 5, 0)),
       plot.subtitle = element_text(colour = "#000000", size = 14,
                                     hjust=0.5, margin = margin(0, 0, 10, 0)),
       plot.caption = element_text(colour = "#000000", size = 9,
                                    hjust=1, margin = margin(10, 0, 20, 0)),
       plot.background = element_rect(fill = "#F4F5F0"),
       panel.background = element_rect(fill = "white",
                                        colour = "#000000", linetype = "solid"),
       panel.grid.major.x = element_line(colour = "gray79", linetype = "dotted"),
       panel.grid.major.y = element_line(colour = "gray79", linetype = "dotted"),
       panel.grid.minor = element_blank(),
        axis.title.x = element_text(size = 10, colour = "#000000", hjust=0.5,
                                    face = "bold", margin = margin(10, 0, 10, 0)),
       axis.title.y = element_text(size = 10, colour = "#000000", face = "bold",
                                    margin = margin(0, 10, 0, 0)),
       axis.text = element_text(size = 10, colour = "#000000"),
       axis.line.y = element_line(colour = "#000000"),
       axis.line.x = element_line(colour = "#000000"),
       axis.ticks = element_line(colour = "#000000", size = 1),
        legend.text = element_text(size = 12, colour = "#000000"),
        legend.background = element_rect(fill = "white", colour = "#000000",
                                         size = 0.3, linetype = "solid"),
        legend.key = element_rect(fill = NA),
        legend.position = "bottom",
        legend.direction = "horizontal")
}
# Trends in life expectancy
italy e0 %>%
  select(Year, Female, Male) %>%
  pivot_longer(., cols = c(Male, Female), names_to = "Sex", values_to = "Life_Exp") %>%
  ggplot(aes(x = Year, y = Life_Exp, color = Sex)) +
  theme_graphs() +
  geom_line(size=1) +
  scale_colour_manual(limits=c("Female", "Male"), values=c("#008c45", "#cd212a")) +
  labs(title = "Evolution of life expectancy at birth",
       subtitle = "Italy, 1872-2017",
```

```
x = "Years", y = "Life Expectancy at Birth",
    caption = "Own elaboration. Source: Human Mortality Database.
    University of California, and Max Planck Institute for Demographic Research.",
    colour = NULL) +
scale_x_continuous(breaks=seq(from=1870,to=2020,by=10),limits=c(1870,2020))
```



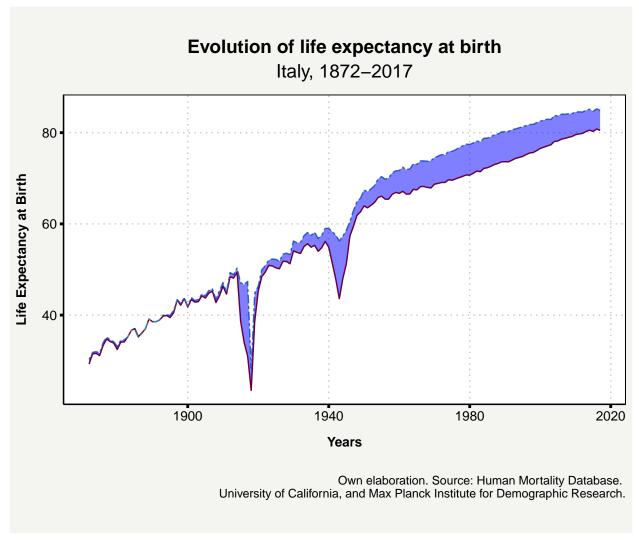
```
ggsave(file="italy_e0.jpeg", width=16, height=8, dpi=300)
```

The life expectancies at birth are increasing both for men and women, although with a sharp decreases in the periods 1914-1918 and 1940-1945 attribuable to war mortality. The life expectancies in 1870 are very similar (30.26 for females and 29.28 for males) but from the XX century the gender gap is becoming more and more pronounced.

# 2c) Visualize the evolution of the gender gap in e0 over time

A first graphical look at the gender gap is showed in the plot below. The gender gap is represented as the area between the curves of the life expectancies at birth of women and men (coloured in violet).

```
italy_e0 %>%
  ggplot(aes(x = Year)) +
  theme_graphs() +
  geom_line(aes(y = Male), color = "darkred") +
  geom_line(aes(y = Female), color="steelblue", linetype="twodash") +
  scale_color_manual(limits=c("Female", "Male"), values=c("#008c45", "#cd212a")) +
  labs(title = "Evolution of life expectancy at birth",
      subtitle = "Italy, 1872-2017",
      x = "Years", y = "Life Expectancy at Birth",
      caption = "Own elaboration. Source: Human Mortality Database.
      University of California, and Max Planck Institute for Demographic Research.",
      colour = NULL) +
  geom_ribbon(aes(ymin = Male, ymax = Female), fill = "blue", alpha = .5)
```

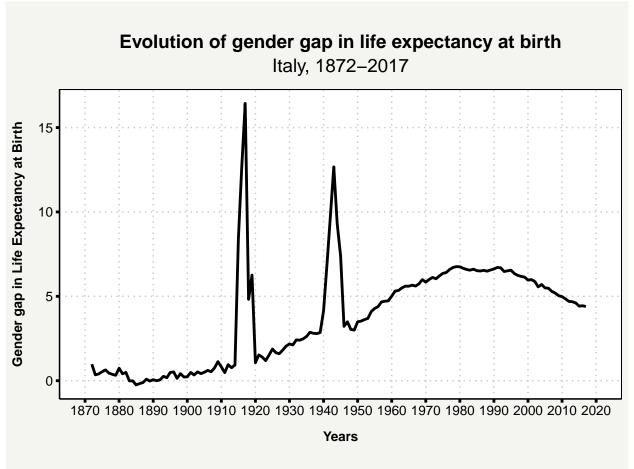


```
ggsave(file="italy_e0_gapA.jpeg", width=16, height=8, dpi=300)
```

The area seems to widen in the periods corresponding to the two World Wars and to become quite constant after the second World War until recent times.

But let's now try to quantify the gender gap in life expectancy at birth as the life expectancy of females minus the life expectancy of males and see how it has developed over time.

```
# Measuring the gender gap
italy_e0$gender_gap <- italy_e0$Female - italy_e0$Male</pre>
# Evolution of the gender gap over time
italy_e0 %>%
  ggplot(aes(x = Year, y = gender_gap)) +
  theme_graphs() +
  geom_line(size=1) +
  scale_colour_manual(limits=c("Female", "Male"), values=c("#008c45", "#cd212a")) +
  labs(title = "Evolution of gender gap in life expectancy at birth",
       subtitle = "Italy, 1872-2017",
       x = "Years",
       y = "Gender gap in Life Expectancy at Birth",
       caption = "Own elaboration. Source: Human Mortality Database.
       University of California, and Max Planck Institute for Demographic Research.",
       colour = NULL) +
  scale_x_continuous(breaks=seq(from=1870,to=2020,by=10),limits=c(1870,2020))
```



Own elaboration. Source: Human Mortality Database. University of California, and Max Planck Institute for Demographic Research.

```
ggsave(file="italy_e0_gapB.jpeg", width=16, height=8, dpi=300)
```

As we observed previously, the gender gap has been increasing from the XX century and was particularly large during the two wars due to the high men mortality. However, we can now notice a deceleration in this increase after the second war world and a closure of the gap starting from the '80s.

# References

Goebel, Jan, Markus M Grabka, Stefan Liebig, Martin Kroh, David Richter, Carsten Schröder, and Jürgen Schupp. 2019. "The German Socio-Economic Panel (Soep)." *Jahrbücher Für Nationalökonomie Und Statistik* 239 (2): 345–60.

HMD. n.d. "Human Mortality Database, University of California, Berkeley (Usa), and Max Planck Institute for Demographic Research (Germany)." https://www.mortality.org.

Riffe, Tim. 2015. "Reading Human Fertility Database and Human Mortality Database Data into R." TR-2015-004. MPIDR. http://www.demogr.mpg.de/en/projects\_publications/publications\_1904/mpidr\_technical\_r eports/reading\_human\_fertility\_database\_and\_human\_mortality\_database\_data\_into\_r\_5438.htm.

Team, R Core, and others. 2013. "R: A Language and Environment for Statistical Computing." Vienna, Austria.