Barcelona Summer School of Demography

Module 2. Demography with R

1. Basic Demographic Measures

Basic Demographic measures

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1 Exercises

1.1 Load the data PopulationSpain.txt and BirthsSpain.txt

```
library(tidyverse)
## -- Attaching packages ------ tidyverse 1.3.1 --
## v ggplot2 3.3.5
                            0.3.4
                   v purrr
## v tibble 3.1.2
                   v dplyr
                            1.0.7
                  v stringr 1.4.0
## v tidyr 1.1.3
                   v forcats 0.5.1
## v readr
         1.4.0
## -- Conflicts ------ tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
library(readr)
#Birth counts in 2014
B <- read_delim("BirthsSpain.txt",</pre>
              delim = " ",
              trim_ws = TRUE,
              skip = 2) \%
```

```
mutate(Age = parse_number(Age)) %>%
      select(Year, Age, Births = Total)
##
## -- Column specification -
## cols(
     Year = col_double(),
##
     Age = col_character(),
##
##
     Total = col_double()
## )
P <- read_delim("PopulationSpain.txt",</pre>
                 delim = " ",
                 trim ws = TRUE,
                 skip = 2,
                 col_types = "ccddd") %>%
      mutate(Age = parse_number(Age)) %>%
      pivot_longer(Female:Total,
                    names_to = "Sex",
                    values_to = "Population") %>%
      arrange(Year, Sex, Age)
```

Note: there was some sort of universe adjustment in the Spanish data in 1975. Possibly a switch from *de facto* to *de jure* population definition. HMD codes such changes to a moment in time, with – suffix on the *before* population and a + suffix on the *after* population. Accounting for this adds some additional logic to the code, but we can pass over that. Actually we could also just remove 1975 altogether if so inclined.

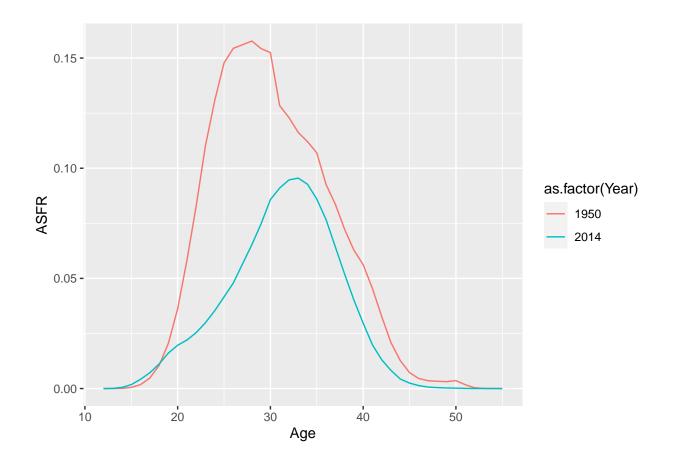
1.2 Calculate age-specific fertility rates from 1950 until 2014 (annually) from age 12 to 55. Skip the year 1975 (if you want to make this easier!).

I'm going to do this the tidy way, first calculating exposure, then merging with births, then calculating rates using mutate(). Two helper functions are nchar(), which tells you the number of characters in string (Year in our case) and grepl() which gives logical character pattern matching (i.e. TRUE if there's a match, otherwise FALSE).

```
Year = Year - 1)
# stick together, put side by side.
# could have also just joined...)
ES <-
P1 %>%
bind_rows(P2) %>%
pivot_wider(names_from = "side", values_from = "Population", names_prefix = "P_") %>%
# calculate exposure
mutate(Exposure = (P_left + P_right) / 2) %>%
# now RIGHT joint to births
# (cuts down exposure to just ages where there is fertility)
right_join(B, by = c("Year", "Age")) %>%
# Calculate ASFR
mutate(ASFR = Births / Exposure)
```

1.3 Plot the age-specific rates in 1950 and 2014.

```
ES %>%
  filter(Year %in% c(1950, 2014)) %>%
  ggplot(aes(x = Age, y = ASFR, color = as.factor(Year), group = Year)) +
  geom_line() +
  scale_color_discrete()
```

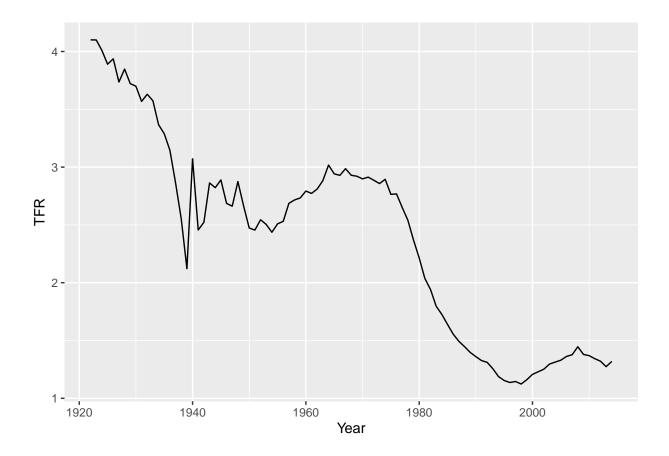


1.4 Calculate the TFR at each year.

```
TFR <-
   ES %>%
   group_by(Year) %>%
   summarize(TFR = sum(ASFR))
```

1.5 Plot TFR over time.

```
TFR %>%
  ggplot(aes(x = Year, y = TFR)) +
  geom_line()
```



1.6 Calculate the mean age at birth over time.

```
MAB <-
ES %>%
mutate(Agemid = Age + .5) %>%
group_by(Year) %>%
summarize(MAB = sum(Agemid * ASFR) / sum(ASFR))
```

1.7 Plot the mean age over time.

```
MAB %>%
  ggplot(aes(x = Year, y = MAB)) +
  geom_line()
```

