

Class Notes

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First merge HMD and HFD

This is a left join, and I figured out which merge function to use by referring to the `dplyr` cheat sheet tiny.cc/37798y and looking at the diagrams on that cheat sheet. Then the final filtering happened basically using demographic logic and not particularly using merge functionality.

```
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.2.1 --
## v ggplot2 3.1.0      v purrr   0.3.2
## v tibble  2.1.1      v dplyr  0.8.0.1
## v tidyr   0.8.3      v stringr 1.4.0
## v readr   1.3.1      v forcats 0.4.0

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()

library(here)

## here() starts at /home/tim/workspace/BSSD/BSSD2019

HMD <- readRDS(here("03_Wednesday", "HMD.rds")) %>%
  filter(Sex == "f")

# now let's get the HFD
HFD <- readRDS("HFD.rds")

# now merge
Stable <- left_join(HMD,
                    HFD,
                    by = c("Country", "Year", "Age")) %>%
  mutate(ASFR = replace_na(ASFR, 0),
         TFR = sum(ASFR)) %>%
  ungroup() %>%
  filter(TFR > 0) %>%
  group_by(Country, Year) %>%
```

Now we want to merge on the SRB. Guess what, if you want info on Male and Female births you look to HMD not HFD. Weird huh? First calculate:

```
# Get info we need about SRB
B <- readRDS("B.rds") %>%
  mutate(SRB = Male / Female,
         PF = Female / Total) %>%
  select(Country, Year, SRB, PF)
```

Now merge:

```
Stable <- left_join(Stable,
  B,
  by = c("Country", "Year"))
```

Now for the sake of getting L_x , we're going to add on all the lifetable columns, using our Wednesday functions, sitting in the file `LifeTableFunctions.R`

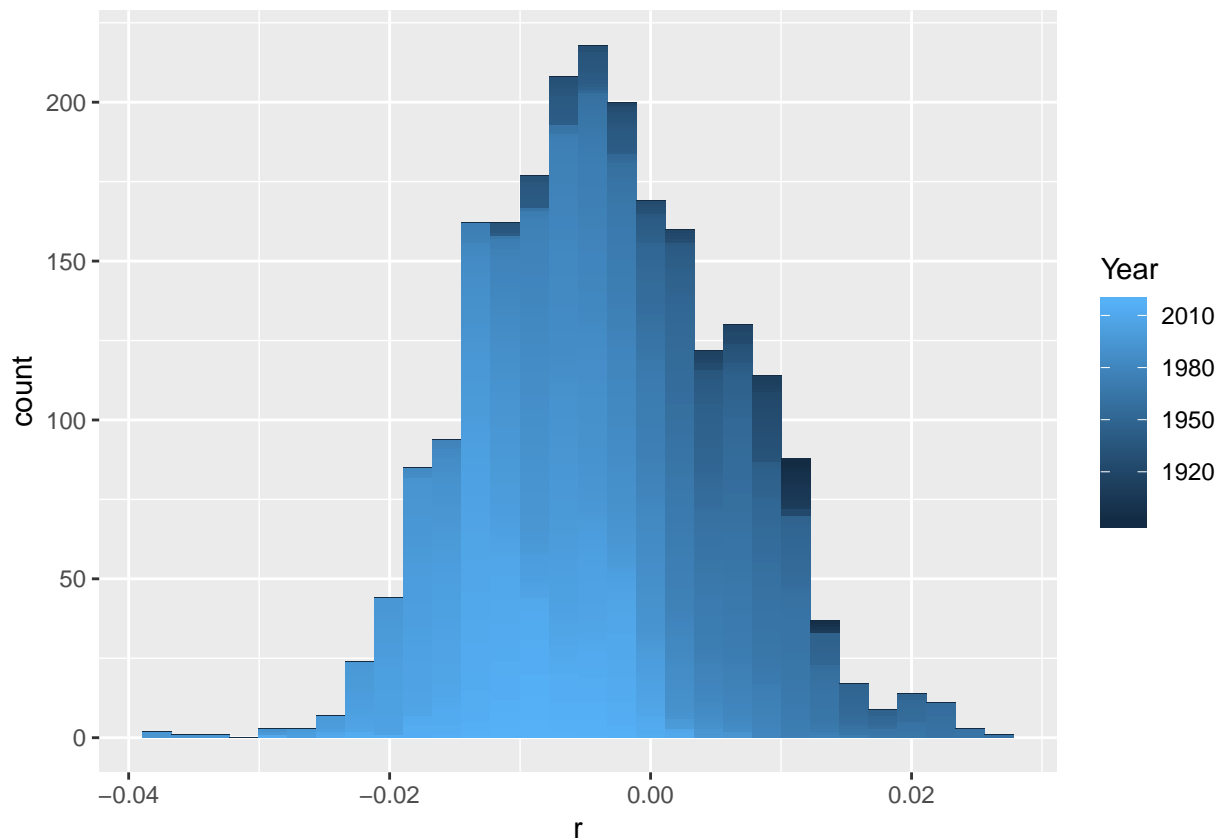
```
source(here("03_Wednesday", "LifeTableFunctions.R"))

Stable <- Stable %>%
  group_by(Country, Year) %>%
  my_lt3()
```

Now we're ready to calculate r , but how do we want to do this?

```
Stable %>%
  group_by(Country, Year) %>%
  mutate(fxf = ASFR * PF) %>%
  summarize(r = Coales_r(Lx, fxf, Age)) %>%
  ggplot(mapping = aes(x = r,
    group = Year,
    fill = Year)) +
  geom_histogram()
```

``stat_bin()`` using ``bins = 30``. Pick better value with ``binwidth``.

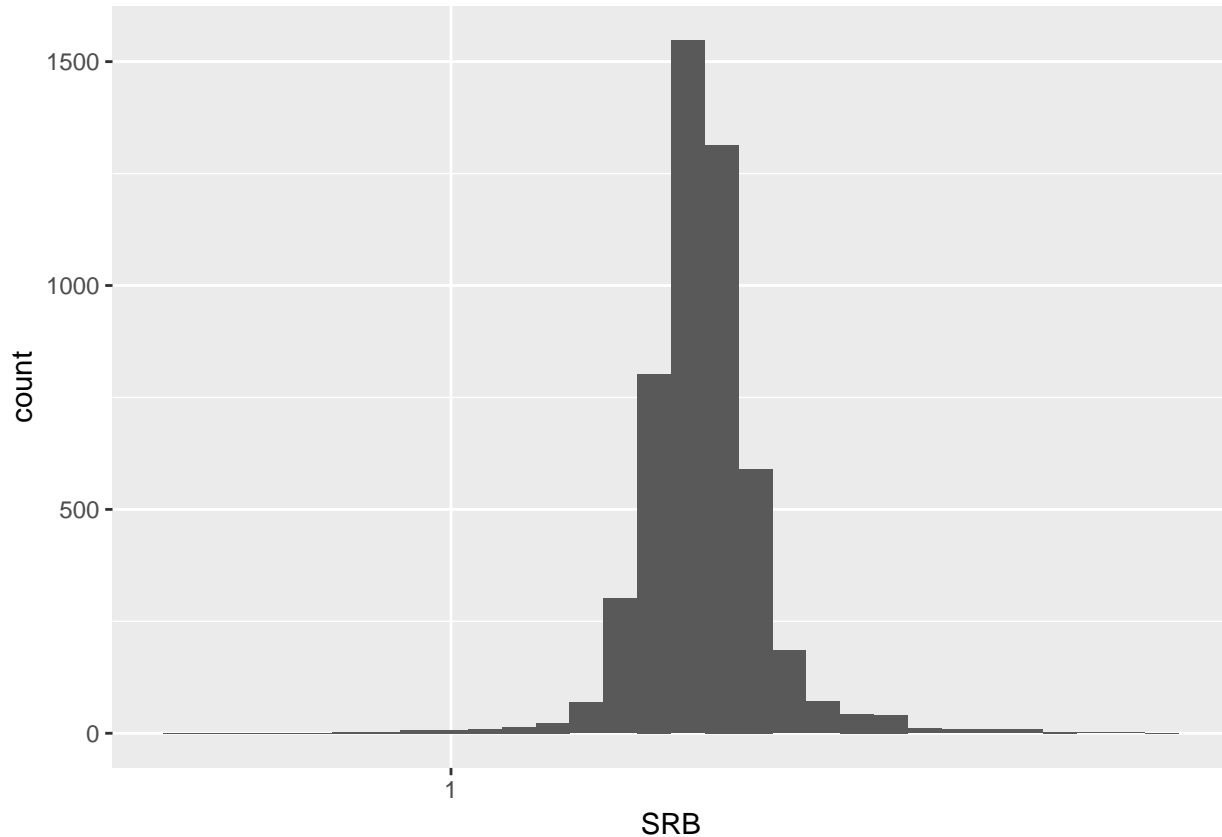


Aside: yes, SRB varies quite a bit:

```
readRDS("B.rds") %>% mutate(SRB = Male / Female) %>%
  ggplot(mapping = aes(x = SRB)) +
  geom_histogram() + scale_x_log10()
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

```
## Warning: Removed 179 rows containing non-finite values (stat_bin).
```



Free exercises:

Here are some variables, some with definitions, possibly new to you, some devised on the spot. Errors fixed along the way, in part along with calibration of calculations.

1. TFR

$$TFR = \sum ASFR(x)$$

2. MAB

$$MAB = \frac{\sum x ASFR(x)}{TFR}$$

3. T (mean generation length in the stable population)

$$T = \frac{\log(R(0))}{r}$$

4. $R(0)$ net reproductive ratio (number of daughters per mother, discounted for mortality and SRB), assumes $L(x)$ has radix = 1, i.e. that $l(0) = 1$

$$R(0) = \sum ASFR(x)^f L(x)$$

5. $e(0)^\dagger$ average years of life lost in the stationary population. Assumes $1 = \sum d(x)$

$$e(0)^\dagger = \sum d(x)e(x)$$

6. net ASFR, often call $\phi(x)$, i.e. ASFR discounted for mortality:

$$\phi(x) = ASFR(x)^f L(x)$$

7. net TFR

$$netTFR = \sum L(x)ASFR(x)$$

8. Replacement TFR, TFR^* , i.e. the quantity that everyone thinks equals 2.1. What is it really? A couple steps. Here derived logically: in the stationary population, $l(0)$ is the number of births. Since, in our version of a stable population the lifetable pertains only to women, we need to produce $l(0) * (SRB + 1)$ births per year in order to achieve stationarity. The number of births we're actually getting is $netTFR$, so we can get a correction factor, k :

$$k = \frac{SRB + 1}{netTFR}$$

Then replacement fertility, TFR^* must be equal to:

$$TFR^* = k * TFR$$

I wonder what it looks like really? Is it higher than 2.1, or lower? Is 2.1 old fashioned?

Some possible calcs + plots

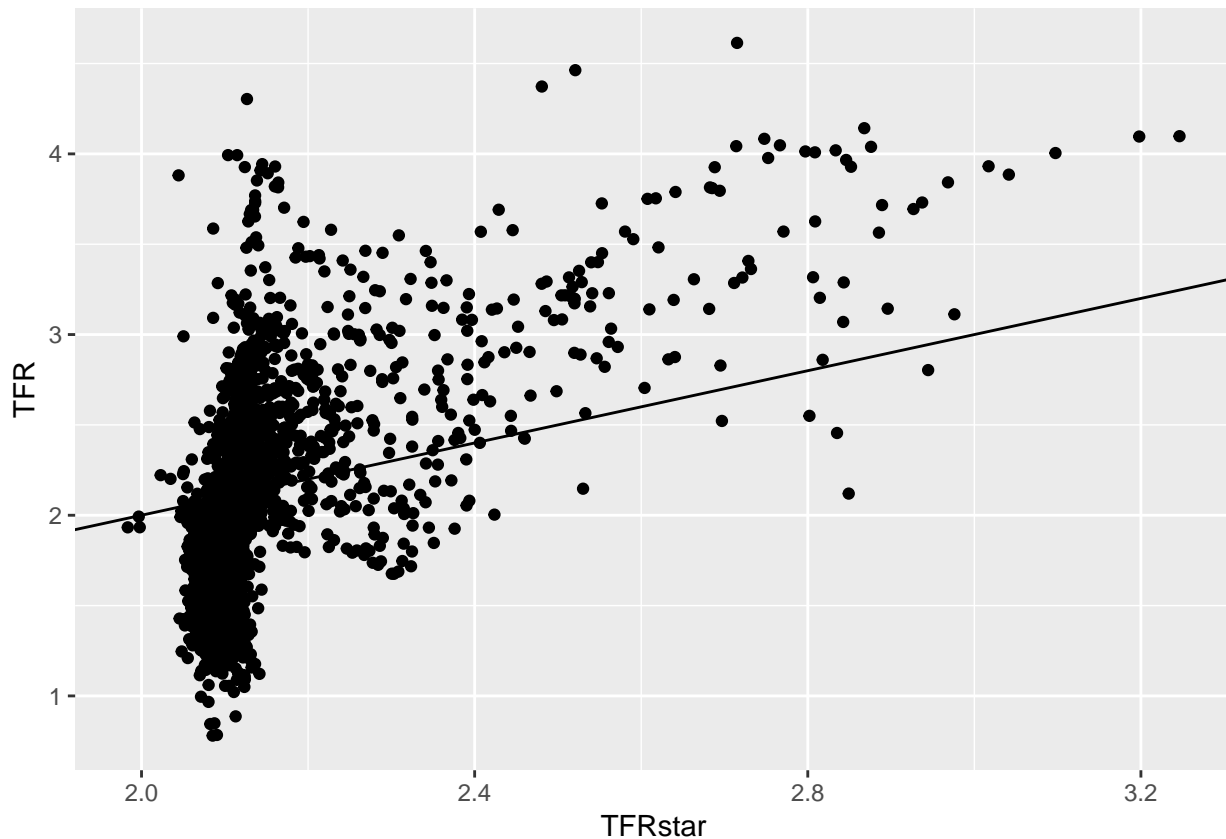
How about TFR^* by ...?

```
# this could be cleaner
Stable2 <- Stable %>%
  group_by(Country, Year) %>%
  mutate(
    fxf = ASFR * PF,      # asfr only females
    phi = fxf * Lx,       # net maternity rates
    R0 = sum(phi),        # NRR net reproductive ratio
    r = Coales_r(Lx, fxf, Age), # intrinsic
    TFR = sum(ASFR),
    netTFR = sum(ASFR * Lx)
  ) %>%
  filter(Age == 0) %>%
  mutate(
    k = (1 + SRB) / netTFR,
    TFRstar = TFR * k,
    TT = log(R0) / r)
```

Below the line is below replacement, above it is above replacement.

```
ggplot(Stable2, mapping = aes(x = TFR, y = TFRstar)) +
  geom_point() + geom_abline(slope=1, intercept=0) +
  coord_fixed() + coord_flip()
```

Coordinate system already present. Adding new coordinate system, which will replace the existing one



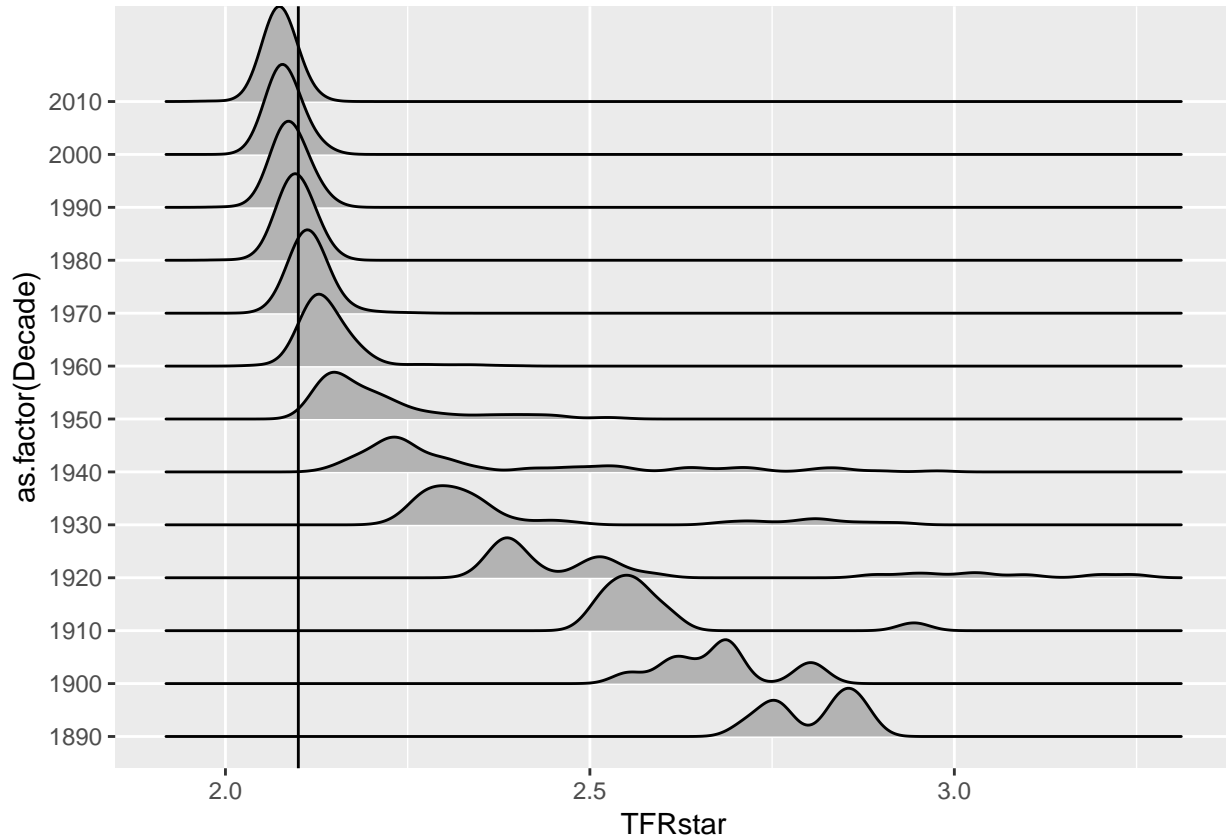
And a final plot in a bit of a hurry, this one was particularly cooperative in its construction. Replacement fertility by decades. Moral of the story holds today though, since the movement is due to mortality: high mortality implies higher TFR needed.

```
library(ggribes)
```

```
##
## Attaching package: 'ggribes'
## The following object is masked from 'package:ggplot2':
##
##   scale_discrete_manual
```

```
Stable2 %>% mutate(Decade = Year - Year %% 10) %>%
ggplot(mapping = aes(x = TFRstar,
                      y = as.factor(Decade))) +
  geom_density_ridges() +
  geom_vline(xintercept=2.1)
```

```
## Picking joint bandwidth of 0.0217
```



And right about here I talked a bunch about motivating work with a clear picture, for purposes of science communication, and gave examples: Oeppen-Vaupel, Mikko, Sander, Kashnitsky :-). And I mentioned some discipline journals that have data viz manuscript types, as well as the above high impact journals where it has worked before for demographers.

Some quantities by age:

(we didn't do anything here together) 1. The stationary age structure is just $L(x)$ 2. The stable age structure, $c(x)$ is:

$$c(x) = \frac{L(x)e^{-rx}}{\sum L(x)e^{-rx}}$$

3. so the mean age of the stationary and stable populations are:

$$A = \frac{\sum xL(x)}{e(0)}$$

$$A^* = \frac{\sum xc(x)}{\sum c(x)}$$

And of course, wouldn't we like to know how A^* changes with r ? Maybe turn it into an elasticity problem?

4. Some people (not me) would take this as an opportunity to introduce dependency ratios, like OADR, the old-age dependency ratio. But I think it's too ill-formed to be useful anymore, so I won't give formulas for that. But there's this not-so-bad alternative from Scherbov & Sanderson, the so-called *prospective* old-age dependency ratio, which is at least benchmarked to mortality conditions.

$$POARD = \frac{\sum P(x|x > x^*)}{P}$$

where:

$$x^* = x \sim e(x) = 15$$

Which is maybe not the best notation, but the idea is to find x where $e(x) = 15$, and then just sum the population up from there. At least it moves with mortality.

Intuition break: stable age structure

This was just for the sake of the picture: we didn't talk about this code really

```
Lx <- Stable %>%
  filter(Country == "USA" &
         Year == 2000) %>%
  pull(Lx)

r <- seq(-.02, .02, length = 21)
Lxr_to_cx <- function(Lx, r, x){
  CX <- Lx * exp(-r * x)
  CX / sum(CX)
}

StablePop <- matrix(nrow = 111,
                    ncol = 21,
                    dimnames = list(Age = 0:110, r = r))
for (i in 1:21){
  StablePop[,i] <- Lxr_to_cx(Lx, r[i], x = 0:110)
}

library(reshape2)

##
## Attaching package: 'reshape2'
## The following object is masked from 'package:tidyr':
##
## smiths

melt(StablePop, varnames = c("Age", "r"),
     value.name = "cx") %>%
  ggplot(mapping = aes(x = Age,
                       y = cx,
                       group = r,
                       color = r)) +
  geom_line()
```

