

# STA2201 Lab 1

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```
library(dplyr)
```

```
##
## Attaching package: 'dplyr'
##
## The following objects are masked from 'package:stats':
##
##   filter, lag
##
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.2 --
## v ggplot2 3.3.6      v purrr 0.3.4
## v tibble 3.1.7       v stringr 1.4.0
## v tidyr 1.2.1        v forcats 0.5.2
## v readr 2.1.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
```

```
library(ggplot2)
```

```
dm <- read_table("https://www.prdh.umontreal.ca/BDLC/data/ont/Mx_1x1.txt", skip = 2, col_types = "dcddd")
```

```
## Warning: 494 parsing failures.
## row    col                expected actual                                file
## 108 Female no trailing characters . 'https://www.prdh.umontreal.ca/BDLC/data/ont/Mx_1x1.txt'
## 109 Female no trailing characters . 'https://www.prdh.umontreal.ca/BDLC/data/ont/Mx_1x1.txt'
## 110 Female no trailing characters . 'https://www.prdh.umontreal.ca/BDLC/data/ont/Mx_1x1.txt'
## 110 Male   no trailing characters . 'https://www.prdh.umontreal.ca/BDLC/data/ont/Mx_1x1.txt'
## 110 Total  no trailing characters . 'https://www.prdh.umontreal.ca/BDLC/data/ont/Mx_1x1.txt'
## ... ..
## See problems(...) for more details.
```

```
head(dm)
```

```
## # A tibble: 6 x 5
##   Year Age   Female   Male   Total
##   <dbl> <chr>   <dbl>   <dbl> <dbl>
## 1  1921 0     0.0978 0.129 0.114
## 2  1921 1     0.0129 0.0144 0.0137
## 3  1921 2     0.00521 0.00737 0.00631
```

```
## 4 1921 3      0.00471 0.00457 0.00464
## 5 1921 4      0.00461 0.00433 0.00447
## 6 1921 5      0.00372 0.00361 0.00367
```

```
dm <- data.frame(dm)
```

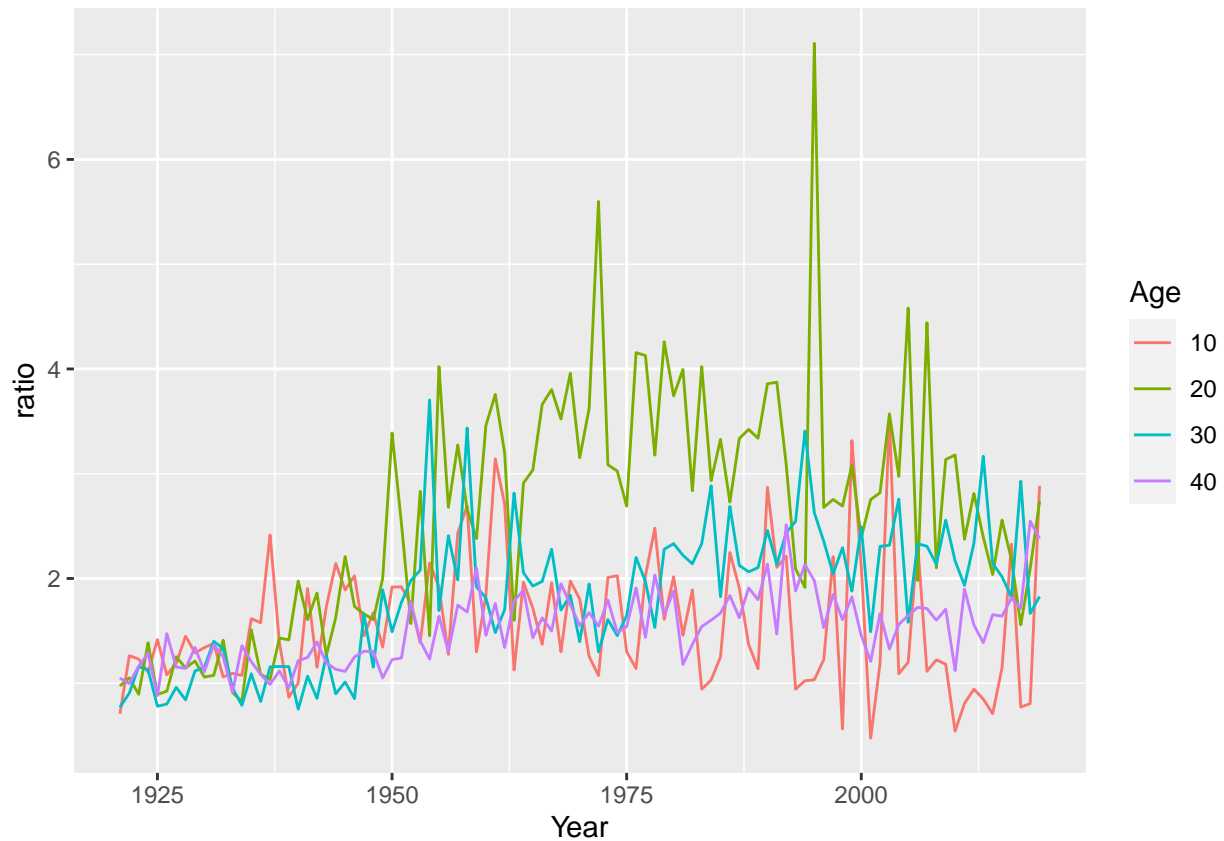
```
pop <- read_table("https://www.prdh.umontreal.ca/BDLC/data/ont/Population.txt", skip = 2, col_types = "d")
head(pop)
```

```
## # A tibble: 6 x 5
##   Year Age  Female  Male  Total
##   <dbl> <chr>  <dbl>  <dbl> <dbl>
## 1 1921 0      30157. 31530. 61687.
## 2 1921 1      30391. 31319. 61711.
## 3 1921 2      30962. 31785. 62747.
## 4 1921 3      31306. 32031. 63336.
## 5 1921 4      31364. 32046. 63409.
## 6 1921 5      31175. 31847. 63021.
```

```
pop <- data.frame(pop)
```

## Q1

```
mort_rates = dm %>%
  filter(
    Age == 10 | Age == 20 | Age == 30 | Age == 40
  ) %>%
  mutate(
    ratio = Male/Female
  )
mort_rates %>%
  ggplot(aes(y = ratio, x = Year, color = Age)) +
  geom_line()
```



```
#plot(mort_rates$ratio, mort_rates$Year, type = "line", color = Age)
```

## Q2

```
max(dm$Female)
```

```
## [1] NA
```

```
# Highest Absolute
```

```
dm %>% group_by(Year) %>% filter(Female == max(Female)) %>% select(Year, Age, Female)
```

```
## # A tibble: 49 x 3
## # Groups:   Year [48]
##   Year Age  Female
##   <dbl> <chr> <dbl>
## 1 1948 99    0.677
## 2 1949 102    0.573
## 3 1950 102    0.629
## 4 1951 110+   2.65
## 5 1953 106    1.16
## 6 1954 110+   4.08
## 7 1955 107    1.02
## 8 1956 110+   4.08
## 9 1957 107    2.08
## 10 1958 110+   4.08
## # ... with 39 more rows
```

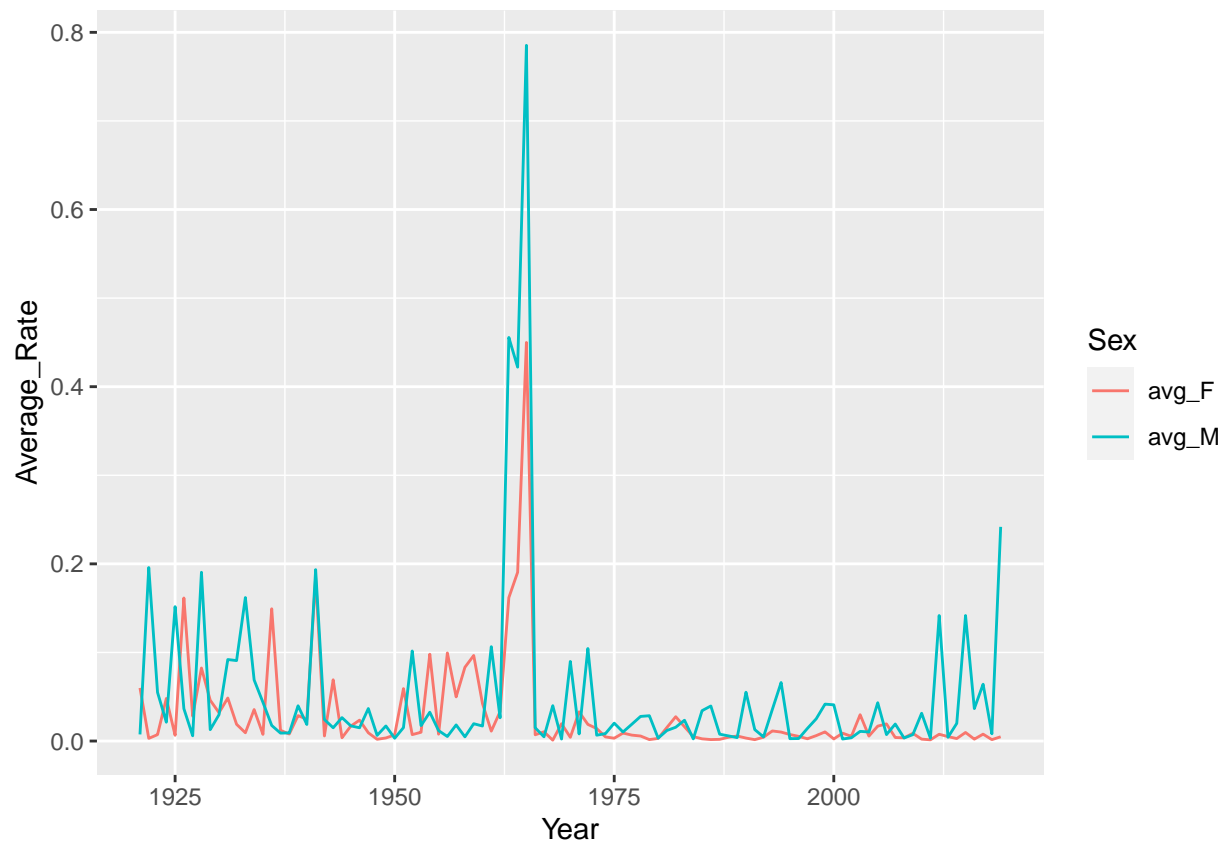
### Q3

```
dm %>%
  group_by(Age) %>%
  summarise(std_M = sd(Male, na.rm = TRUE),
            std_F = sd(Female, na.rm = TRUE),
            std_T = sd(Total, na.rm = TRUE))
```

```
## # A tibble: 111 x 4
##   Age      std_M      std_F      std_T
##   <chr>    <dbl>    <dbl>    <dbl>
## 1 0      0.0330    0.0256    0.0294
## 2 1      0.00396  0.00352    0.00374
## 3 10     0.000561  0.000474  0.000509
## 4 100    0.138      0.0928    0.0729
## 5 101    0.158      0.125     0.0995
## 6 102    0.214      0.143     0.114
## 7 103    0.371      0.252     0.208
## 8 104    1.01       0.449     0.363
## 9 105    1.29      1.27      1.27
## 10 106    1.13      1.21      1.20
## # ... with 101 more rows
```

### Q4

```
dm %>% left_join(y = pop, by = c("Year", "Age")) %>% mutate(
  pw_F = Female.x/Female.y,
  pw_M = Male.x/Male.y
) %>% filter(!is.nan(pw_M) & !is.nan(pw_F) &
            !is.infinite(pw_M) & !is.infinite(pw_F)) %>%
  group_by(Year) %>%
  summarise(avg_F = mean(pw_F, na.rm = TRUE), avg_M = mean(pw_M, na.rm = TRUE)) %>%
  pivot_longer(cols = c(2:3), names_to = "Sex", values_to = "Average_Rate") %>%
  ggplot(aes(x = Year, y = Average_Rate, color = Sex)) + geom_line()
```



From the looks of this graph, men tend to see a much higher variability in their death rates, often attaining greater levels than Women.