## PS 138Z - The Politics of Immigration: Section 7

2024-03-7

#### Introduction

Today, we will continue using ggplot to analyze data from the Mexican Migration Project (MMP) Database to produce.

### Cleaning data

First, let's load the data using the read.csv() function. This time, we will assign the dataset to an object called mmp.data.

```
mmp.data <- read.csv("mmp_subset.csv")</pre>
```

Let's upload tidyverse package only. You can upload a (pre-installed) package with the library() function. library(tidyverse)

```
## Warning: package 'tibble' was built under R version 4.2.3
```

In this exercise, we will work with four variables: "Currently on last migration" (usyrl), "age" (age), "sex" (sex), and "education years" (edyrs). Remember that you can see the description of these and other variables in the dataset in the codebook. Let's start by creating a dataset with only these variables. Complete the code below:

Now, let's examine if any of these four variables has any numerical missing values. You can do this in multiple ways. This time, we will use the range() function. Complete the code below (Hint: remember that you should use the argument na.rm = TRUE because we do not know if the variables have missing values coded as NA as well):

```
range(no.miss.data$uscurtrp, na.rm = TRUE)

## [1]    1 8888

range(no.miss.data$age, na.rm = TRUE)

## [1]    0 8888

range(no.miss.data$sex, na.rm = TRUE)

## [1]    1 2

range(no.miss.data$edyrs, na.rm = TRUE)
```

## [1] 0 8888

What can you conclude? Which variables have numerical missing values?

Remember that for some functions, such as mean() and range(), we can ignore missing values coded as NA with the argument na.rm = TRUE. Also, remember that we can easily drop all the rows with at least one NA using the function omit.na(). For these reasons, we may prefer to replace the numerical missing values with NAs. Complete the code below:

```
no.miss.data$uscurtrp[no.miss.data$uscurtrp == 8888] <- NA
no.miss.data$age[no.miss.data$age == 8888] <- NA
no.miss.data$edyrs[no.miss.data$edyrs == 8888] <- NA
```

Now that we have replaced all the numerical missing values with NAs, we can drop any rows with NAs.

```
no.miss.data <- na.omit(no.miss.data)
```

To this point, we have a "clean" dataset with only the variables we are interested in and no missing values. As a final step, let's recode the sex and uscurtrp variables to facilitate plotting. Run the code below:

```
no.miss.data$uscurtrp[no.miss.data$uscurtrp == 1] <- "Yes"
no.miss.data$uscurtrp[no.miss.data$uscurtrp == 2] <- "No"

no.miss.data$sex[no.miss.data$sex == 1] <- "Male"
no.miss.data$sex[no.miss.data$sex == 2] <- "Female"</pre>
```

Use the head() function to look at the first 15 rows in the dataset.

```
head(no.miss.data, 15)
```

```
##
       uscurtrp age
                         sex edyrs
## 1
                  26
                       Male
                                  9
            Yes
## 7
             No
                  59
                       Male
                                  6
## 8
                  27
                       Male
                                  9
            Yes
## 12
             No
                  52
                        Male
                                  0
                  27
                                  9
## 15
            Yes
                       Male
## 18
             No
                  55
                        Male
## 20
            Yes
                  31
                       Male
                                  9
## 25
                  40
                        Male
            Yes
                                 11
## 26
            Yes
                  41 Female
                                  4
                                  7
## 31
            Yes
                  26
                        Male
## 35
             No
                  30
                                  6
                       Male
## 36
             No
                  34
                       Male
                                  6
## 38
                  30
                       Male
                                  6
            Yes
## 39
            Yes
                  30 Female
                                  9
                                  3
## 40
            Yes
                   8
                       Male
```

## Characterizing recent immigrants

Complete the code below to calculate the average age and educational level by immigration status.

```
no.miss.data %>%
  group_by(uscurtrp) %>%
  summarise_at(vars(age, edyrs), c(mean), na.rm = TRUE)
```

Complete the code below to cross-tabulate immigration status and sex.

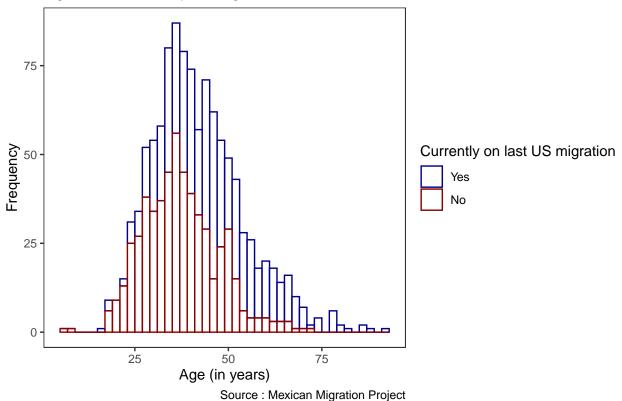
#### table(no.miss.data\$sex, no.miss.data\$uscurtrp)

What can you conclude from these descriptive statistics?

Now, let's use the ggplot package to plot a grouped histogram showing the age distribution by immigration status. Complete the code below:

```
plot1 <- ggplot(no.miss.data, aes(x=age, color= factor(uscurtrp))) +
   geom_histogram(binwidth = 2, fill="white", alpha=0.5) +
   labs(
     title = "Age Distribution by Immigration Status",
     caption = "Source : Mexican Migration Project",
     x = "Age (in years)",
     y = "Frequency",
     color = "Currently on last US migration") +
   scale_color_manual(labels = c("Yes", "No"), values = c("darkblue", "darkred")) +
   theme_test()
plot1</pre>
```

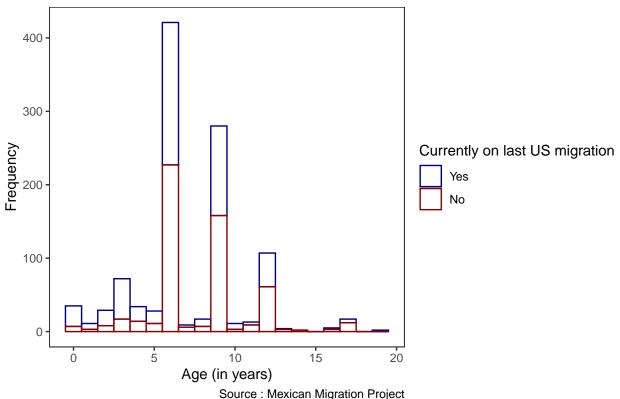
## Age Distribution by Immigration Status



Similarly, let's use the ggplot package to plot a grouped histogram showing the distribution of school years completed by immigration status. Complete the code below:

```
plot2 <- ggplot(no.miss.data, aes(x=edyrs, color= factor(uscurtrp))) +
   geom_histogram(binwidth = 1, fill="white", alpha=0.5) +
   labs(
     title = "Age Distribution by Immigration Status",
     caption = "Source : Mexican Migration Project",
     x = "Age (in years)",
     y = "Frequency",
     color = "Currently on last US migration") +
   scale_color_manual(labels = c("Yes", "No"), values = c("darkblue", "darkred")) +
   theme_test()
plot2</pre>
```

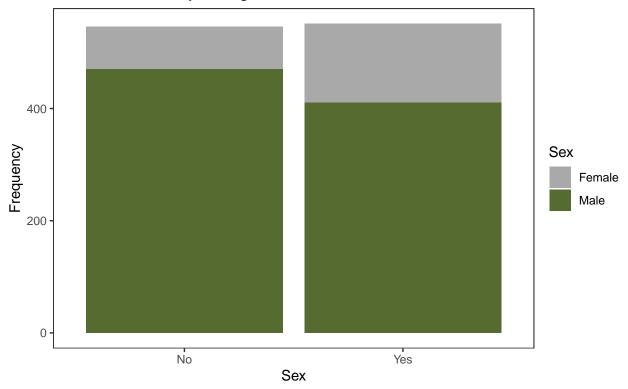
## Age Distribution by Immigration Status



Finally, let's use <code>geom\_bar()</code> to create a grouped bar plot showing how individuals with different immigration statuses are distributed by sex. Complete the code below:

plot2

# Sex Distribution by Immigration Status



Source: Mexican Migration Project

How do the descriptive tables and figures we produced relate to the theories we have discussed in class?