Introduction to R and RStudio

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IMMERSE Project



The Institute of Mixture Modeling for Equity-Oriented Researchers, Scholars, and Educators (IMMERSE) is an IES funded training grant (R305B220021) to support Education scholars in integrating mixture modeling into their research.

- Please visit our website to learn more and apply for the year-long fellowship.
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How to reference this walk through: *This work was supported by the IMMERSE Project* (IES - 305B220021) Visit our GitHub account to download the materials needed for this walk through.

Introduction to R and RStudio

This walk through is presented by the IMMERSE team and will go through some common tasks carried out in R. There are many free resources available to get started with R and RS tudio. One of our favorites is R for Data Science.

PART 1: Installation

Step 0: Install R, RStudio, and Mplus

Here you will find a guide to installing both R and R Studio. You can also install Mplus here.

Note: The installation of Mplus requires a paid license with the mixture add-on. IMMERSE fellows will be given their own copy of Mplus for use during the one year training.

PART 2: Set-up

Step 1: Create a new R-project in RStudio

R-projects help us organize our folders , filepaths, and scripts. To create a new R project:

• File \rightarrow New Project...

Click "New Directory" -> New Project -> Name your project

Step 2: Create an R-markdown document

An R-markdown file provides an authoring framework for data science that allows us to organize our reports using texts and code chunks. This document you are reading was made using R-markdown!

To create an R-markdown:

• File \rightarrow New File \rightarrow R Markdown...

In the window that pops up, give the R-markdown a title such as "Introduction to R and RStudio" Click "OK." You should see a new markdown with some example text and code chunks. We want a clean document to start off with so delete everything from line 10 down. Go ahead and save this document in your R Project folder.

Table 1: LCA Indicators

Name	Label	Values
leaid	District Identification Code	
ncessch	School Identification Code	
report_dis	Number of students harassed or bullied on the basis of disability	0 = No reported incidents, $1 = At$ least one reported incident
report_race	Number of students harassed or bullied on the basis of race, color, or national origin	0 = No reported incidents, $1 = At$ least one reported incident
report_sex	Number of students harassed or bullied on the basis of sex	0 = No reported incidents, $1 = At$ least one reported incident
counselors_fte	Number of full time equivalent counselors hired as school staff	0 = No staff present, $1 = At$ least one staff present
report_sex	Number of full time equivalent psychologists hired as school staff	0 = No staff present, $1 = At$ least one staff present
$counselors_fte$	Number of full time equivalent law enforcement officers hired as school staff	0 = No staff present, $1 =$ At least one staff present

Step 3: Load packages

Your first code chunk in any given markdown should be the packages you will be using. To insert a code chunk, etiher use the keyboard shortcut ctrl + alt + i or Code -> Insert Chunk or click the green box with the letter C on it. There are a few packages we want our markdown to read in:

```
library(psych) # describe()
library(here) #helps with filepaths
library(gt) # create tables
library(tidyverse) #collection of R packages designed for data science
```

As a reminder, if a function does not work and you receive an error like this: could not find function "random_function"; or if you try to load a package and you receive an error like this: there is no package called `random_package`, then you will need to install the package using install.packages("random_package") in the console (the bottom-left window in R studio). Once you have installed the package you will *never* need to install it again, however you must *always* load in the packages at the beginning of your R markdown using library(random_package), as shown in this document.

The style of code and package we will be using is called **tidyverse**. Most functions are within the **tidyverse** package and if not, I've indicated the packages used in the code chunk above.

PART 3: Explore the data

Step 4: Read in data

To demonstrate mixture modeling in the training program and online resource components of the IES grant we utilize the *Civil Rights Data Collection (CRDC)* (CRDC) data repository. The CRDC is a federally mandated school-level data collection effort that occurs every other year. This public data is currently available for selected latent class indicators across 4 years (2011, 2013, 2015, 2017) and all US states. In this example, we use the Arizona state sample. We utilize six focal indicators which constitute the latent class model in our example; three variables which report on harassment/bullying in schools based on disability, race, or sex, and three variables on full-time equivalent school staff hires (counselor, psychologist, law enforcement). This data source also includes covariates on a variety of subjects and distal outcomes reported in 2018 such as math/reading assessments and graduation rates.

To read in data in R:

```
data <- read_csv(here("data", "crdc_lca_data.csv")) %>%
  mutate_if(is.character, as.numeric)
# Ways to view data in R:
# 1. click on the data in your Global Environment (upper right pane) or use...
View(data)
# 2. summary() gives basic summary statistics & shows number of NA values
# *great for checking that data has been read in correctly*
summary(data)
```

##	leaid	ncess	ch	report_di	s repor	t_race
##	Min. :4000	001 Min. :	4.000e+10	Min. :0.0	000 Min.	:0.000
##	1st Qu.:4008	304 1st Qu.:	4.008e+10	1st Qu.:0.0	000 1st Qu	ı.:0.000
##	Median :4034	120 Median :	4.034e+10	Median :0.0	000 Median	:0.000
##	Mean :4038	365 Mean :	4.038e+10	Mean :0.04	425 Mean	:0.103
##	3rd Qu.:4063	330 3rd Qu.:	4.063e+10	3rd Qu.:0.0	000 3rd Qu	ı.:0.000
##	Max. :4097	734 Max. :	4.097e+10	Max. :1.0	000 Max.	:1.000
##	NA's :20	NA's :	40	NA's :27	NA's	:27
##	report_sex	counselors	_fte ps	ych_fte	law_fte	
##	Min. :0.00) Min. :0.	0000 Min.	:0.0000	Min. :0.0	0000
##	1st Qu.:0.00) 1st Qu.:0.	0000 1st	Qu.:0.0000	1st Qu.:0.0	0000
##	Median :0.00) Median :0.	0000 Medi	an :0.0000	Median :0.0	0000
##	Mean :0.17	7 Mean :0.	4595 Mean	:0.4742	Mean :0.1	.255
##	3rd Qu.:0.00) 3rd Qu.:1.	0000 3rd	Qu.:1.0000	3rd Qu.:0.0	0000
##	Max. :1.00) Max. :1.	0000 Max.	:1.0000	Max. :1.0	0000
##	NA's :27	NA's :27	NA's	:30	NA's :27	

3. names() provides a list of column names. Very useful if you don't have them memorized! names(data)

##	[1]	"leaid"	"ncessch"	"report_dis"	"report_race"
##	[5]	"report_sex"	"counselors_fte"	"psych_fte"	"law_fte"

4. head() prints the top x rows of the dataframe head(data)

##	#	A tibbl	Le: 6 x 8					
##		leaid	ncessch	report_dis	report_race	report_sex	counselors_fte	psych_fte
##		<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>
##	1	400001	40000100120	0	0	0	1	1
##	2	400001	40000100616	0	0	1	1	1
##	3	400001	40000101204	0	0	1	1	1
##	4	400001	40000101871	0	1	1	1	1
##	5	400001	40000101872	0	0	0	1	1
##	6	400001	40000102344	0	0	0	1	1
##	#	wit	ch 1 more van	riable: law	fte <dbl></dbl>			

Step 5: Descriptive Statistics

Let's look at descriptive statistics for each variable. Because looking at the ID variables' (leaid) and (necessch) descriptives is unnecessary, we use select() to remove the variable by using the minus (-) sign:

```
data %>%
  select(-leaid, -ncessch) %>%
  summary()
```

##	repor	t_dis	report	_race	repor	t_sex	counsel	ors_fte
##	Min.	:0.0000	Min.	:0.000	Min.	:0.00	Min.	:0.0000
##	1st Qu.	:0.0000	1st Qu.	:0.000	1st Qu.	:0.00	1st Qu.	:0.0000
##	Median	:0.0000	Median	:0.000	Median	:0.00	Median	:0.0000
##	Mean	:0.0425	Mean	:0.103	Mean	:0.17	Mean	:0.4595
##	3rd Qu.	:0.0000	3rd Qu.	:0.000	3rd Qu.	:0.00	3rd Qu.	:1.0000
##	Max.	:1.0000	Max.	:1.000	Max.	:1.00	Max.	:1.0000
##	NA's	:27	NA's	:27	NA's	:27	NA's	:27
##	psych	n_fte	law_	fte				
##	Min.	:0.0000	Min.	:0.0000				
##	1st Qu.	:0.0000	1st Qu.	:0.0000				
##	Median	:0.0000	Median	:0.0000				
##	Mean	:0.4742	Mean	:0.1255				
##	3rd Qu.	:1.0000	3rd Qu.	:0.0000				
##	Max.	:1.0000	Max.	:1.0000				
##	NA's	:30	NA's	:27				

Alternatively, we can use the psych::describe() function to give more information:

```
data %>%
  select(-leaid, -ncessch) %>%
  describe()
```

leaid

##		vars	n	\mathtt{mean}	sd	median	trimmed	\mathtt{mad}	min	\max	range	skew
##	report_dis	1	2000	0.04	0.20	0	0.00	0	0	1	1	4.53
##	report_race	2	2000	0.10	0.30	0	0.00	0	0	1	1	2.61
##	report_sex	3	2000	0.17	0.38	0	0.09	0	0	1	1	1.76
##	counselors_fte	4	2000	0.46	0.50	0	0.45	0	0	1	1	0.16
##	psych_fte	5	1997	0.47	0.50	0	0.47	0	0	1	1	0.10
##	law_fte	6	2000	0.13	0.33	0	0.03	0	0	1	1	2.26
##		kurto	osis	se								
##	report_dis	18	3.55 (0.00								
##	report_race	4	1.82 (0.01								
##	report_sex	:	1.08 (0.01								
##	$\verb"counselors_fte"$	-:	1.97 (0.01								
##	psych_fte	-:	1.99 (0.01								
##	law_fte	3	3.11 (0.01								

What if we want to look at a subset of the data? For example, what if we want to subset the data to observe a specific school district? (leaid) We can use tidyverse::filter() to subset the data using certain criteria.

```
data %>%
filter(leaid == 408800) %>%
describe()
## vars n mean sd median trimmed mad
```

1 86 4.088e+05

0.00 4.088e+05 4.088e+05 0.0 4.088e+05

min

##	ncessch	2 86 4	.088e+10	493.16	4.088e+10	0 4.088e+10	89.7	4.088e+10
##	report_dis	3 86 5	.000e-02	0.21	0.000e+00	0.000e+00	0.0	0.000e+00
##	report_race	4 86 1	.500e-01	0.36	0.000e+00	0 7.000e-02	0.0	0.000e+00
##	report_sex	5 86 1	.900e-01	0.39	0.000e+0	0 1.100e-01	0.0	0.000e+00
##	counselors_fte	6869	.500e-01	0.21	1.000e+0	0 1.000e+00	0.0	0.000e+00
##	psych_fte	7 86 1	.900e-01	0.39	0.000e+0	0 1.100e-01	0.0	0.000e+00
##	law_fte	8 86 1	.400e-01	0.35	0.000e+0	0 6.000e-02	0.0	0.000e+00
##		max	range	skew ku	rtosis	se		
##	leaid	4.088e+05	0	NaN	NaN O	.00		
##	ncessch	4.088e+10	2597	2.58	7.77 53	.18		
##	report_dis	1.000e+00	1 4	4.23	16.10 0	.02		
##	report_race	1.000e+00	1	1.91	1.68 0	.04		
##	report_sex	1.000e+00	1	1.59	0.52 0	.04		
##	$\verb"counselors_fte"$	1.000e+00	1 -	4.23	16.10 0	.02		
##	psych_fte	1.000e+00	1	1.59	0.52 0	.04		
##	law_fte	1.000e+00	1 1	2.04	2.21 0	.04		
#Yo	ou can use any d	operator t	o filter	: >, <,	==, >=,	etc.		

References

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