## Exam3

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1. Clear the environment. [5 points]

```
#clear environment
rm(list=ls(all=TRUE))
  2. Use the tidycensus package to
 (a) find the inequality Gini index variable explained on the last exam
library(tidycensus)
## Warning: package 'tidycensus' was built under R version 4.0.2
library(tidyverse)
## -- Attaching packages ------ tidyverse 1.3.
## v ggplot2 3.3.1 v purrr
                             0.3.4
## v tibble 3.0.1 v dplyr
                             1.0.0
## v tidyr 1.1.0 v stringr 1.4.0
## v readr 1.3.1 v forcats 0.5.0
## -- Conflicts ----- tidyverse_conflicts(
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                 masks stats::lag()
#getting access to census and acs
census_api_key("b0f75887f4369253f295f4c59d8527a5d420cf32",
             install = TRUE, overwrite = TRUE)
## Your original .Renviron will be backed up and stored in your R HOME directory if needed.
## Your API key has been stored in your .Renviron and can be accessed by Sys.getenv("CENSUS_API_KEY").
## To use now, restart R or run 'readRenviron("~/.Renviron")'
```

## [1] "b0f75887f4369253f295f4c59d8527a5d420cf32"

```
#loading 2010 and 2015 variables
acs2010 = load_variables(year = 2010, "acs5")
acs2015 <- load_variables(year = 2015, "acs5")</pre>
```

(b) import in the state-level inequality Gini estimates for 2010 and 2015 in the five-year American Community Survey as a single panel dataset;

## Getting data from the 2006-2010 5-year ACS

## Getting data from the 2011-2015 5-year ACS

```
import_gini_2015$year = 2015
#append data
import_gini = bind_rows(import_gini_2010,import_gini_2015)
```

(c) rename estimate as gini in your final data frame, which you should call inequality\_panel;

```
##
## Attaching package: 'data.table'

## The following objects are masked from 'package:dplyr':
##
## between, first, last

## The following object is masked from 'package:purrr':
##
## transpose
```

(d) rename NAME to state as well;

inequality\_panel = setnames(import\_gini, "estimate", "gini")

library(data.table)

```
#rename
setnames(inequality_panel,"NAME","state")
```

- (e) ensure that inequality\_panel has a year variable so we can distinguish between the 2010 and 2015 gini index data;
- Done in step b.
  - (f) as a final step, run the head() command so we can get a quick peak at inequality\_panel (Hint: you may need to import each year separately and then append the two data frames together.) [15 points]

## head(inequality\_panel)

```
## # A tibble: 6 x 6
##
     GEOID state
                      variable gini
                                       moe year
                               <dbl> <dbl> <dbl>
     <chr> <chr>
                      <chr>>
## 1 01
           Alabama
                      gini
                               0.47 0.003 2010
## 2 02
           Alaska
                      gini
                               0.412 0.006
                                            2010
## 3 04
                               0.453 0.002 2010
           Arizona
                      gini
## 4 05
           Arkansas
                               0.459 0.003
                                             2010
                      gini
## 5 06
           California gini
                               0.469 0.001
                                             2010
## 6 08
                               0.455 0.003 2010
           Colorado
                      gini
```

3. Reshape the inequality\_panel wide, such that the gini values for 2010 and 2015 have their own columns. Also, please keep both the state and GEOID variables. Call the resulting data frame inequality\_wide. After you are done with the reshape, run the head() command so we can get a quick peak at the data. [5 points]

```
## # A tibble: 6 x 6
##
     GEOID state
                       variable gini
                                             year
                                        moe
##
     <chr> <chr>
                       <chr>
                                <dbl> <dbl> <dbl>
## 1 01
           Alabama
                      gini
                                0.47 0.003
                                             2010
## 2 02
           Alaska
                                0.412 0.006
                                             2010
                      gini
## 3 04
           Arizona
                      gini
                                0.453 0.002
                                             2010
## 4 05
                                0.459 0.003
                                             2010
           Arkansas
                      gini
## 5 06
           California gini
                                0.469 0.001
                                             2010
## 6 08
                                0.455 0.003
                                             2010
           Colorado
                      gini
```

4. Reshape inequality\_wide to long format. Once you are done, run the head() command so we can get a quick peak at the data. [5 points]

```
#pivoting longer
inequality_long =
inequality_wide %>%
 pivot_longer(cols = starts_with("year"),
             names_to = "year",
             names_prefix = "year_",
             values_to = "gini",
             values_drop_na = FALSE)
#head
head(inequality_long)
## # A tibble: 6 x 4
    state GEOID year
                        gini
    <chr> <chr> <chr> <chr> <dbl>
##
## 1 Alabama 01
                  2010 0.47
## 2 Alabama 01
                  2015 0.475
## 3 Alaska 02
                  2010 0.412
## 4 Alaska 02
                  2015 0.418
## 5 Arizona 04
                  2010 0.453
## 6 Arizona 04
                  2015 0.465
  5. Show with some R code that inequality_panel and inequality_long have the same number of observa-
    tions. [5 points]
#number of observations for long
str(inequality_long)
## tibble [104 x 4] (S3: tbl_df/tbl/data.frame)
## $ state: chr [1:104] "Alabama" "Alabama" "Alaska" "Alaska" ...
## $ GEOID: chr [1:104] "01" "01" "02" "02" ...
## $ year : chr [1:104] "2010" "2015" "2010" "2015" ...
## $ gini : num [1:104] 0.47 0.475 0.412 0.418 0.453 ...
dim(inequality_long)
## [1] 104
#number of observations for panel
str(inequality_panel)
## tibble [104 x 6] (S3: tbl_df/tbl/data.frame)
## $ GEOID : chr [1:104] "01" "02" "04" "05" ...
            : chr [1:104] "Alabama" "Alaska" "Arizona" "Arkansas" ...
## $ state
## $ variable: chr [1:104] "gini" "gini" "gini" "gini" ...
## $ gini : num [1:104] 0.47 0.412 0.453 0.459 0.469 0.455 0.482 0.436 0.535 0.471 ...
             : num [1:104] 0.003 0.006 0.002 0.003 0.001 0.003 0.003 0.005 0.005 0.002 ...
## $ moe
             $ year
dim(inequality_panel)
```

```
## [1] 104 6
```

They both have 104 observations, but different variable numbers.

6. Collapse the inequality\_long data frame by state, such that you obtain a single mean gini score for each state for the years 2010 and 2015. When collapsing, also keep both the GEOID and state variables. Call your resulting data frame inequality\_collapsed. [5 points]

```
#collapsing, summarizing. summarizing numeric in unique state
inequality_collapsed =
inequality_long %>%
  group_by(state,GEOID) %>%
  summarize(across(where(is.numeric),sum))
```

- ## 'summarise()' regrouping output by 'state' (override with '.groups' argument)
  - 7. Produce a map of the United States that colors in the state polygons by their mean gini scores from inequality\_collapsed, using the WGS84 coordinate system. When doing so, use the viridis color scheme. (Note: there are a few different ways to produce the map. We will accept any one of these ways, and the answer key will detail 3 ways. If you want to choose the method with the shape file, you can get the state-level shape file on the Census page). [10 points]

```
library(rio)
library(tidyverse)
library(googlesheets4)

## Warning: package 'googlesheets4' was built under R version 4.0.2

library(labelled)

## Warning: package 'labelled' was built under R version 4.0.2

library(data.table)
library(varhandle)
library(ggrepel)

## Warning: package 'ggrepel' was built under R version 4.0.2

library(geosphere)

## Warning: package 'geosphere' was built under R version 4.0.2

library(rgeos)
```

## Warning: package 'rgeos' was built under R version 4.0.2

## Loading required package: sp

```
## Warning: package 'sp' was built under R version 4.0.2
## rgeos version: 0.5-3, (SVN revision 634)
## GEOS runtime version: 3.8.0-CAPI-1.13.1
## Linking to sp version: 1.4-2
## Polygon checking: TRUE
library(viridis)
## Loading required package: viridisLite
#library(mapview)
library(rnaturalearth)
## Warning: package 'rnaturalearth' was built under R version 4.0.2
library(devtools)
## Warning: package 'devtools' was built under R version 4.0.2
## Loading required package: usethis
## Warning: package 'usethis' was built under R version 4.0.2
library(rnaturalearthhires)
library(sp)
library(sf)
## Warning: package 'sf' was built under R version 4.0.2
## Linking to GEOS 3.8.0, GDAL 3.0.4, PROJ 6.3.1
library(ggplot2)
#get map of US
#america <- ne_countries(country = 'United States',</pre>
 # scale = "medium",
  #returnclass = "sf")
\#usa\_map = ggplot() +
 \# geom\_sf(data = america) +
  #geom_sf(data = inequality_collapsed, aes(fill = 'gini'))+
  #scale_fill_viridis(option = "viridis")+
  #qqtitle("")+
  #theme(plot.title = element_text(hjust=0.5)) +
  #theme_void()
#print(usa_map)
```

8. Use the WDI package to import in data on Gross Domestic Product (GDP) in current US dollars. When doing so, include all countries and only the years 2006 and 2007. Rename your GDP variable to gdp\_current. [5 points]

9. Deflate gdp\_current to constant 2010 or 2015 US dollars, and call the new variable gdp\_deflated. In words, also tell us the base year that you picked and why. At the end, run a head() command to prove that everything works. [5 points]

10. In a Shiny app, what are the three main components and their subcomponents? [5 points]

```
#library(shiny)
#component 1: User Interface (UI) [an HTML page]
    #subcomponent: assigning input and outputs
#ui <- fluidPage("Hello, World")
#component 2: server function
    #subcomponent: rendering functions, assigning it to output variables
#server <- function(input,output) {
#}
#component 3: execution, telling shiny what the ui and server is.
#shinyApp(ui = ui,server = server)</pre>
```

11. Pull this .pdf file from Mike Denly's webpage. It is a report on governance in Armenia that Mike Denly and Mike Findley prepared for the US Agency for International Development (USAID). [5 points]

```
library(pdftools)
```

## Warning: package 'pdftools' was built under R version 4.0.2

```
## Using poppler version 0.73.0
```

```
library(tidyr)
library(tidytext)
```

## Warning: package 'tidytext' was built under R version 4.0.2

```
library(dplyr)
library(stringr)
#importing pdf
md_import = pdf_text(pdf = "https://pdf.usaid.gov/pdf_docs/PAOOTNMG.pdf")
```

12. Convert the text pulled from this .pdf file to a data frame, using the , stringsAsFactors=FALSE option. Call the data frame armeniatext. [5 points]

```
#convert to data frame
armeniatext = data.frame(md_import,stringAsFactors=FALSE)
```

13. Tokenize the data by word and then remove stop words. [5 points]

```
data("stop_words")
#tokenized and removed stop words
armeniatext = armeniatext %>%
   unnest_tokens(word,md_import) %>%
   anti_join(stop_words)
```

```
## Joining, by = "word"
```

14. Figure out the top 5 most used word in the report. [5 points]

```
#reordering by count
armeniatext %>%
  count(word, sort = TRUE) %>%
  group_by(n) %>%
  top_n(n=5)
```

## Selecting by n

```
## # A tibble: 4,228 x 2
## # Groups: n [74]
##
     word
                    n
##
      <chr>
                <int>
## 1 armenia
                  252
## 2 political
                  207
## 3 corruption
                  186
## 4 governance
                  185
## 5 democracy
                  132
## 6 evidence
                  126
## 7 review
                  121
## 8 service
                  119
                  106
## 9 2017
## 10 public
                  105
## # ... with 4,218 more rows
```

15. Load the Billboard Hot 100 webpage, which we explored in the course modules. Name the list object: hot100exam [5 points]

```
## Loading required package: xml2
##
## Attaching package: 'rvest'
## The following object is masked from 'package:purrr':
##
## pluck
## The following object is masked from 'package:readr':
##
## guess_encoding
#read html
hot100 = "https://www.billboard.com/charts/hot-100"
```

16. Use rvest to obtain identify all of the nodes in the webpage. [5 points]

hot100exam = read\_html(hot100)

```
body_nodes <- hot100exam %>%
  html_node("body") %>%
  html_children()
```

17. Use Google Chrome developer to identify the necessary tags and pull the data on Rank, Artist, Title, and Last Week. HINT 1: In class we showed you how to get the first three of these. You simply need to add the Last Week ranking. HINT 2: You can navigate two ways. Hovering to find what you need or by doing Cmd+F / Ctrl+F and using actual data to find the location. HINT 3: You're looking to update the code based on the way the information is in referenced. Try out some different options and see what shows up in the environment. Keep trying until you see that you have a chr [1:100] with values that correspond to what is in the web page. [5 points]

Final question. Save all of the files (i.e. .Rmd, .dta, .pdf/Word Doc), push them to your GitHub repo, and provide us with the link to that repo. [no points; 15-point penalty for lack of submission (see above)] exam3 github