Coding Lab Final Project

Harris Coding Camp

Summer 2024

Your final project, which is optional, is quite simple. You will pick a data set that speaks to you and try to uncover something interesting which you will visualize in a plot. You will also compute some summary statistics that you will show in a summary table. Your TAs will review your work.

Due date: September 30, 12pm upload your Rmd and pdf to Gradescope. Feel free to submit your work earlier. We will not review work that is turned in after that point.

Submission: Your project should be uploaded to Gradescope, a function embedded on the Canvas website. It won't accept HTML files for the project, so you must submit a pdf instead. To do so, simply render your Rmd as a pdf directly and submit that file.

Format:

You will turn in a knitted pdf that has the following sections.

- graph in which you load your data set and provide the minimal code that produces your graph.
- table in which you create and print a summary table with minimal code.
- appendix (optional) in which you share code you used during data exploration, e.g. extensions of your main plot, other plots you attempted on your search for your main plot. (Think about this as a place to share code you want feedback on or that help further your story.)

We have provided a sample project at the end of this document.

You are welcome to use Google and Stack Overflow as you proceed. Please cite your sources if you borrow code from Stack Overflow or someone's blog. To cite, just add a comment with the url. See, the last code line of the sample project where we used Stack Overflow to figure out how to reformat the legend.

We will review how to use Rmds with you. But here are some quick tips.

- You make a new section in Rmd using # section title.
- When you read in you data, we do not want to see messages or warnings. To avoid this start the code block where you read the data with the following {r, message = FALSE, warning = FALSE}.
- If you have a line of code that is too long, it will be cutoff. Most R code can be split across two lines.
- Knit early and often! This is how you know if the Rmd is working how you think it is.
- To make your table look nice, you can use knitr::kable(your_data). (See example below).

Datasets:

Below is a list of suitable data sources. You are welcome to and encouraged to find a data source not on this list that speaks to your policy interests. Many of these data sources have a wide range of data sets. Pick

one that comes in tabular format with several variables that are interesting¹. We recommend that once you pick a data source that is sufficient stick with it, so you have ample time to focus on your R skills.

¹By which it means there's variation.

Description	url
Weekly Covid data from US CDC (several	https://www.cdc.gov/nchs/nvss/vsrr/covid_weekly/
datasets available, follow links)	
Washington Post fatal police shooting data	https://github.com/washingtonpost/data-police-shootings
has records of every fatal shooting in the	
United States by a police officer in the line of	
duty since Jan. 1, 2015. Their github has	
other data mixed in with code they use for	
other stuff.	
Open Policing has traffic stop data for several	https://openpolicing.stanford.edu/data/
police departments with varying time	
horizons and variables	
Eviction Lab has eviction data at the block	https://evictionlab.org/get-the-data/
group / tract level from $2000-2016$	
Google maps data aggregating how visits to	https://www.google.com/covid19/mobility/index.html?hl=en
places, such as grocery stores and parks, are	
changing in each geographic region since	
February 15, 2020 until today (3-4 days	
delay), compared to the same week of the day	
in January, 2020.	
World Inequality Database which allows you	https://wid.world/data/
to download a customizable dataset. You are	
able to choose the indicators you want (per	
adult gdp, top 10% income share & dozens	
others), countries you want and date range	
that you want.	
The Humanitarian Data Exchange (HDX) is	https://data.humdata.org/
an open platform for sharing data across	
crises and organizations. They host	
thousands of datasets including development	
indicator data, geospatial data, damage	
assessments, and more.	
World Bank publishes hundreds of different	https://data.worldbank.org/
global development related datasets including	
datasets on World Development Indicators,	
all of World Bank's lending projects and	
access to sample survey data etc. Also able	
to search data by country or indicator.	
The City of Chicago publishes many different	https://data.cityofchicago.org/
datasets, including ones on public finance,	
public safety, transportation, and education	https://opendata.cityofpoyworl- us/
NYC also has an open data initiative that	https://opendata.cityofnewyork.us/
aims to provide data from different agencies	
in one central platform. Data on ride-share	
programs can be found there as well, but of	
course, the public version	https://www.leagsle.com/date/177699
List of datasets related to black lives and	https://www.kaggle.com/data/177628
police violence. Kaggle is a platform for	
learning data science through competitions.	

FiveThirtyEight is a website focused on	https://data.fivethirtyeight.com/
opinion poll analysis, politics, economics, and	
sports blogging. Currently, it has a unique	
repository of datasets about 2020 election	
polls and forecasts, Trump's popularity,	
Americans' view on COVID crisis Trump's	
response, NBA/NFL/MLB/Soccer	
predictions, among others.	
Propublica is a investigative news	https://www.propublica.org/datastore/
organization that does in-depth data	
reporting. Browse data sets about Health,	
Criminal Justice, Education, Politics,	
Business, Transportation, Military,	
Environment, Finance, or Religion. Not all	
data here is free.	
U.S. Bureau of Labor Statistics	https://www.bls.gov/
Bureau of Economic Analysis	https://www.bea.gov/
Federal Reserve Economic Data	https://fred.stlouisfed.org/
The University of Michigan Health and	https://hrs.isr.umich.edu/
Retirement Study (HRS)	
The Panel Study of Income Dynamics (PSID)	https://psidonline.isr.umich.edu/
National Bureau of Economic Research has a	https://data.nber.org/data/
master page of data resources.	
Google also has a dataset search tool that is	https://datasetsearch.research.google.com/
pretty handy for finding data without having	
to comb through niche websites. I would not	
rely on this completely, though, since not all	
datasets follow the standards required to	
show up in these search results	

Example project:

Introduction

I analyze weekly covid-19 data from the US Center for Disease Control. I show the extent to which racial disparities exist as measured by the percentage change in deaths in 2020 compared to 2015-2019. The plot below shows data for the United States except the tri-state area NY-NJ-CT. NYC is a large diverse city that was particularly hard hit by the coronavirus, so it is plausible that the racial disparities reported on are driven by those facts. The plot shows that NYC does not appear to drive the disparities. In the appendix, I examine the same question in states with high Latinx populations that experienced a covid-19 surge in the late summer (TX-CA-AZ-FL). And, I look at the least densely populated states.

The data description is found here: $https://www.cdc.gov/nchs/nvss/vsrr/covid_weekly/ \ The \ data \ can be \ downloaded \ directly \ from \ here: \ https://data.cdc.gov/api/views/qfhf-uhaa/rows.csv?accessType= DOWNLOAD&bom=true&format=true%20target=$

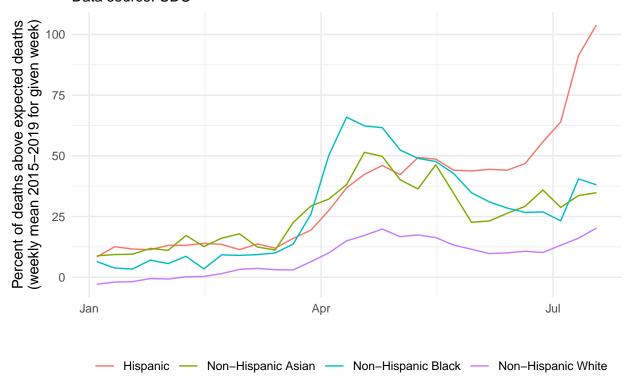
graph

```
library(tidyverse)
covid data <-
  read_csv("Weekly_counts_of_deaths_by_jurisdiction_and_race_and_Hispanic_origin.csv",
           col_types = cols(Suppress = col_character())) %>%
   mutate(week = `Week Ending Date`,
           race_ethnicity = `Race/Ethnicity`,
           n_deaths = `Number of Deaths`,
           diff = `Difference from 2015-2019 to 2020`,
           expected_deaths = n_deaths - diff,
           perc diff = `Percent Difference from 2015-2019 to 2020`,
           year = MMWRYear,
           week_no = MMWRWeek,
           jurisdiction = Jurisdiction,
           state = `State Abbreviation`
           ) %>%
   filter(`Time Period` == "2020", Outcome == "All Cause", Type != "Unweighted") %%
    select(jurisdiction, state, week, year, week_no,
           race_ethnicity, n_deaths, expected_deaths, diff, perc_diff)
```

'summarise()' has grouped output by 'race_ethnicity'. You can override using
the '.groups' argument.

```
data_for_plot %>%
  ggplot(aes(x = week, color = race_ethnicity)) +
    geom_line(aes(y = perc_above_expected)) +
  theme_minimal() +
  labs(y = "Percent of deaths above expected deaths\n(weekly mean 2015-2019 for given week)",
        x = "",
        title = "Racial disparities of Covid-19, USA excluding NY-NJ-CT" ,
        subtitle = "Data source: CDC",
        color = "") +
  theme(legend.position = "bottom")
```

Racial disparities of Covid–19, USA excluding NY–NJ–CT Data source: CDC



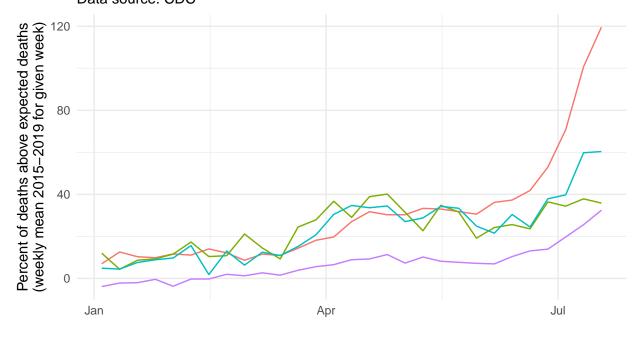
table

race_ethnicity	expected_deaths to	otal_additional_deaths	percent_diff
Hispanic	121787	54977	45.141928
Non-Hispanic American Indian or Alaska	10749	2335	21.722951
Native			
Non-Hispanic Asian	40741	14364	35.256867
Non-Hispanic Black	204431	61969	30.312917
Non-Hispanic White	1337231	130967	9.793895
Other	15094	2613	17.311515

Appendix

```
# We don't require you to use functions. For this project it's acceptable to repeat code.
# As you grow as a programmer, when you find that you want to copy and paste a code block
# over and over again. It means it's time for a function or a loop.
data_for_plot <-
  function(states,
    ethnicities = c("Hispanic", "Non-Hispanic White", "Non-Hispanic Black", "Non-Hispanic Asian")) {
    covid_data %>%
      mutate(week = lubridate::mdy(week)) %>%
      filter(race_ethnicity %in% ethnicities) %>%
      filter(! state %in% "US", state %in% states, week_no <= 29) %>%
      group_by(race_ethnicity, week ) %>%
      summarize(actual_deaths = sum(n_deaths, na.rm = TRUE),
                diff_deaths = sum(diff, na.rm = TRUE),
                expected_deaths = actual_deaths - diff_deaths,
                perc above expected = 100 * diff deaths / expected deaths)
}
make_plot <- function(data_for_plot, title) {</pre>
   data_for_plot %>%
      ggplot(aes(x = week, color = race_ethnicity)) +
        geom_line(aes(y = perc_above_expected)) +
      theme_minimal() +
      labs(y = "Percent of deaths above expected deaths\n(weekly mean 2015-2019 for given week)",
           x = "",
           title = glue::glue("Racial disparities of Covid-19 {title}"),
           subtitle = "Data source: CDC",
           color = "") +
      theme(legend.position = "bottom")
}
data_for_plot(c("AZ", "TX", "FL", "CA")) %>% make_plot("in TX-FL-AZ-CA")
## 'summarise()' has grouped output by 'race_ethnicity'. You can override using
## the '.groups' argument.
```

Racial disparities of Covid–19 in TX–FL–AZ–CA Data source: CDC



Hispanic — Non-Hispanic Asian — Non-Hispanic Black — Non-Hispanic White

```
## 'summarise()' has grouped output by 'race_ethnicity'. You can override using
## the '.groups' argument.
```

Warning: Removed 29 rows containing missing values or values outside the scale range
('geom_line()').

