

Using R effectively

1 October 2021

EARNCon 2021

Ben Zipperer

Economic Policy Institute

bzipperer@epi.org

[@benzipperer](https://twitter.com/benzipperer)

https://economic.github.io/data_bootcamp/

1. Review last time

- American Community Survey data
- Low-wage workers in Virginia

2. Recoding new variables with “if” conditions

3. Combining and transforming datasets: stacking, joining, and reshaping

4. Complex analysis: county-level statistics using ACS

5. Project management in Rstudio

Last time: review, data cleaning

```
# Load the ACS data from IPUMS
acs <- read_dta("/home/benzipperer/Downloads/acs_2019.dta")

# Clean up the data
acs_clean <- acs %>%
  # keep only workers
  filter(incwage > 0 & incwage < 999998) %>%
  filter(uhrswork > 0) %>%
  # full-year workers only
  filter(wkswork2 == 6) %>%
  # restrict analysis to VA
  filter(statefip == 51) %>%
  # define wages and low-wage workers
  mutate(wage = incwage / (uhrswork * 51)) %>%
  mutate(low_wage = wage <= 15)
```

```
# Shares of low-wage workers, overall and by demographic cuts
```

```
acs_clean %>%
```

```
  summarize(weighted.mean(low_wage, perwt))
```

```
acs_clean %>%
```

```
  group_by(sex) %>%
```

```
  summarize(weighted.mean(low_wage, perwt))
```

```
acs_clean %>%
```

```
  group_by(race) %>%
```

```
  summarize(weighted.mean(low_wage, perwt))
```

Specific tasks

1. Redefine race category to identify Hispanics
 - a. define indicator for Hispanic ethnicity/origin
 - b. redefine “race” to be more coarse and include Hispanic origin
2. Expand analysis to use all workers rather than just full-year

Examples

1. define 0-1 Hispanic ethnicity from detailed country of origin
2. define aggregated race variable from detailed race
3. define “average” weeks worked, based on binned weeks worked

Useful functions

- *ifelse()*
- *case_when()*

Useful functions

- `ifelse(test, yes, no)` creates values=yes/no corresponding to test=true/false
- `case_when(test1 ~ value1, test2 ~ value2, ...)`
assigns value if test true

For complex recoding, *always* double-check the results

- something like `count(oldvar, newvar)` can be very helpful

Specific tasks

1. Create single summary dataset with race-specific *and* overall shares of low-wage workers
2. Add more summary statistics: population counts and sample sizes
3. How does VA compare to the US overall?
4. How does VA compare to nearby states?

Useful functions

- `bind_rows()`, `is.na()`
- `summarize()` summary functions `sum()` and `n()`
- `rename()` and `full_join()`
- multiple groups in `group_by()`
- `pivot_wider()`

ACS data contains one substate identifier

- *puma* = PUMA or Public-Use Microdata Area
- PUMAs are state-specific
- but can overlap several counties

Construct and join PUMA -> county mapping to ACS data

- Geocorr: <http://mcdc.missouri.edu/applications/geocorr2018.html>
- re-scale sample weights to account for PUMA -> county duplication

Do not automatically save/restart your workspace

R projects

Directories