

Data Appendix to “Name of your paper”

Iris Zhong

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1 Raw data

1.1 Gwinnett County 2019 Referendum Dataset

Citation: Results—Gwinnett—Election Night Reporting. (n.d.). Retrieved March 11, 2020, from <https://results.enr.clarityelections.com/GA/Gwinnett/94961/Web02.225391/#/>

DOI: N/A

Date Downloaded: Mar 11, 2020

Filename(s): raw_data/vote_result.xls *If you have a large number of files you can use a patten (see visit data below)* **Unit of observation:** Precinct **Dates covered:** Mar 19, 2019

1.1.1 To obtain a copy

Users can visit the website that displays election results at Gwinnett County at <https://results.enr.clarityelections.com/GA/Gwinnett/94961/Web02.225391/#/> and choose the **Detail XLS** link at the bottom right corner.

The xls file contains three spreadsheets. I will be using the second and third sheet.

1.1.2 Importable version

Filename(s): importable_data/vote_result_importable.xlsx

The raw dataset is hard to be imported to R directly because of the following reasons. First, it has three tabs. Second, the top two rows do not contain any useful information or should be incorporated to the next row. The file uses the extension .xls, which is incompatible with R. Therefore, an importable version of the dataset was created.

Here are the steps:

1. The original file was opened in Microsoft Excel.
2. The turnout rate for each precinct which was displayed in the second spreadsheet was moved to the third spreadsheet.
3. The first and second spreadsheets were deleted.
4. The top two rows of the third spreadsheet were removed.
5. The columns were renamed to reflect whether the vote was for or against the proposal.
6. The extension was changed from .xls to .xlsx.

1.1.3 Variable descriptions

I cannot find any information that describes the variables in this dataset. Therefore, the description below is my understanding.

- **precinct:** The name of the precinct.
- **registered_voters:** The number of registered voters in the precinct.
- **total_votes:** The number of votes received in this referendum.
- **voter_turnout:** The percentage of voters who voted in this referendum. (*registered_voters/total_votes*)
- **election_day_yes:** The number of people who voted yes during the election day.
- **absentee_mail_yes:** The number of people who voted yes by mailing paper ballots prior to the election day.
- **advance_in_person_1_yes:** The number of people who voted yes prior to the election day.
- **advance_in_person_2_yes:** The number of people who voted yes prior to the election day. The difference between this variable from the previous one is not clear. My speculation is they record people voting on different days before election.
- **provisional_yes:** The number of people who voted yes but had questions in their eligibility.
- **votes_yes:** The number of people who voted yes in total. (the sum of the previous five variables)
- **election_day_no:** The number of people who voted no during the election day.
- **absentee_mail_no:** The number of people who voted no by mailing paper ballots prior to the election day.
- **advance_in_person_1_no:** The number of people who voted no prior to the election day.
- **advance_in_person_2_no:** The number of people who voted no prior to the election day. The difference between this variable from the previous one is not clear. My speculation is they record people voting on different days before election.
- **provisional_no:** The number of people who voted no but had questions in their eligibility.
- **votes_no:** The number of people who voted no in total. (the sum of the previous five variables)

1.1.4 Data import code and summary

```
vote_result <- read_excel("importable_data/vote_result_importable.xlsx",
  col_types = c("text", "numeric", "numeric",
    "numeric", "numeric", "numeric",
    "numeric", "numeric", "numeric", "numeric",
    "numeric", "numeric", "numeric", "numeric", "numeric",
    "numeric"))
```

```
View(vote_result)
export_summary_table(dfSummary(vote_result))
```

1.2 Gwinnett County Census Data

Citation: U.S. Census Bureau. (2018). American Community Survey. <https://data.census.gov/cedsci/table?d=ACS%205-Year%20Estimates%20Data%20Profiles&table=DP05&tid=ACSDP5Y2018.DP05&g=0500000US13135,13135.140000&hidePreview=true&t=Age%20and%20Sex%3AHousing%3AHousing%20Units%3ARace%20and%20Ethnicity>

DOI: N/A

Date Downloaded: Mar 19, 2020

Filename(s): N/A

Unit of observation: Census tract

Dates covered: 2018 (5-year estimate)

1.2.1 To obtain a copy

Users can obtain a copy of the dataset from an R package `tidycensus`.

Citation: Walker, K., Eberwein, K., & Herman, M. (2020). `tidycensus`: Load US Census Boundary and Attribute Data as “tidyverse” and ‘sf’-Ready Data Frames (Version 0.9.6) [Computer software]. <https://CRAN.R-project.org/package=tidycensus>

Below are the steps to use `tidycensus` to obtain the data: 1. Install the package `tidycensus` in R.

2. If haven’t, register to get an API key in order to download data from the package. The key can be acquired at http://api.census.gov/data/key_signup.html.

3. Load the key into R with the following code:

`census_api_key(key, install = TRUE)` 4. Load the library in R. Execute the code chunk below to get the dataframe `acs18`. `acs18` shows all of the variables present in ACS-5 2018 survey and their IDs.

```
library(tidycensus)
```

```
## Warning: package 'tidycensus' was built under R version 3.5.3
```

```
acs18 <- load_variables(2018, "acs5", cache = TRUE)
```

5. Search for desirable variables in `acs18` and record their IDs. The selected variables are: population, median income, median age, white population, the number of people who work, the number of people who commute by car, the number of people who commute by public transportation, the number of people who commute by subway, the number of people who go to work by bike, the number of people who walk to work, the number of people who use other transportation means, and the number of people who work at home. Subway is a subcategory of public transportation, but since it is particularly important in this project, it is also selected. Besides, variables that reflect workers’ travel time to work are potentially useful.

```
cbdata <- get_acs(geography = "tract",
  variables = c(total = "B01001_001",
    medincome = "B19013_001",
    medage = "B01002_001",
    white = "B01001A_001",
    transportation_total = "B08006_001",
    car = "B08006_002",
    public = "B08006_008",
    subway = "B08006_011",
    bike = "B08006_014",
    walk = "B08006_015",
    other_transport = "B08006_016",
```

```

no_transport = "B08006_017",
time_total = "B08012_001",
time_less_5 = "B08012_002",
time_5_9 = "B08012_003",
time_10_14 = "B08012_004",
time_15_19 = "B08012_005",
time_20_24 = "B08012_006",
time_25_29 = "B08012_007",
time_30_34 = "B08012_008",
time_35_39 = "B08012_009",
time_40_44 = "B08012_010",
time_45_59 = "B08012_011",
time_60_89 = "B08012_012",
time_more_90 = "B08012_013"),

state = "GA",
county = "Gwinnett",
year = 2018)

```

Getting data from the 2014–2018 5-year ACS

`View(cbddata)`

The code above constructs a dataframe called `cbddata` by calling the function `get_acs()`, which pulls the data from the American Community Survey. Inside the function, the unit of measurement is specified by the `geography` argument. In this case, select “tract” for census tract. Put all the chosen variables in the `variables` argument. Finally, address `state` (“GA”), `county` (“Gwinnett”), and `year` of survey (“2018”).

1.2.2 Importable version (Data Wrangling)

The current dataframe requires modifications. First, each row displays one variable from one tract. However, to make tract as the unit of measurement, each row should include all the variables of one tract. Second, instead of the actual number of people who are white, the percentage of white population is more informative. Similarly, the percentage of people who go to work by certain transportation should also be calculated. Finally, travel time data is mostly grouped by a 5-minute band, which is too detailed for this project. It will be categorized with wider range to reduce variables. After consolidation, the percentages will be calculated.

Here are the steps of data wrangling:

1. Remove the margin of error for each measurement (`moe`), because it is not useful in later analyses.

```

cbddata_moe <- cbddata %>%
  select (-moe)

```

2. Use `pivot_wider()` to transpose the data.

```

cbddata_wider <- cbddata_moe %>%
  pivot_wider(names_from = variable,
              values_from = estimate)

```

3. Calculate the percentage of white population (*white/total*).

```

cbddata_white <- cbddata_wider %>%
  mutate(white_pct = white / total) %>%
  select (-white)

```

4. Calculate the percentage of people using each transportation method.

```

cbddata_transport <- cbddata_white %>%
  mutate(car_pct = car / transportation_total,

```

```

public_pct = public / transportation_total,
subway_pct = subway / transportation_total,
bike_pct = bike / transportation_total,
walk_pct = walk / transportation_total,
other_pct = other_transport / transportation_total,
no_pct = no_transport / transportation_total) %>%
select(-c(car, public, subway, bike, walk, other_transport, no_transport))

```

5. Combine the ranges of travel time data and calculate the percentages.

```

cbdata_tidy <- cbdata_transport %>%
mutate(time_0_29_pct = (time_less_5 + time_10_14 + time_15_19 +
                        time_20_24 + time_25_29) / time_total,
       time_30_59_pct = (time_30_34 + time_35_39 + time_40_44 +
                        time_45_59) / time_total,
       time_60_89_pct = time_60_89 / time_total,
       time_more_90_pct = time_more_90 / time_total) %>%
select(-c(time_less_5, time_5_9, time_10_14, time_15_19, time_20_24,
          time_25_29, time_30_34, time_35_39, time_40_44, time_45_59,
          time_60_89, time_more_90))

```

1.2.3 Variable descriptions

- **GEOID:** The geographic identifier of the census tract.
- **NAME:** The name of the census tract.
- **total:** The total population of the tract.
- **medage:** The median age of the population in the tract.
- **medincome:** The median income of the population in the tract.
- **white_pct:** The percentage of white population in the tract.
- **transportation_total:** The number of people who were sampled in the transportation survey.
- **car_pct:** The percentage of people who go to work by car, truck or van.
- **public_pct:** The percentage of people who go to work by public transportation (excluding taxi or cab).
- **subway_pct:** The percentage of people who go to work by subway or elevated.
- **bike_pct:** The percentage of people who go to work by bike.
- **walk_pct:** The percentage of people who go to work on foot.
- **other_pct:** The percentage of people who go to work by other transportation means such as taxi, cab and motorcycle.
- **no_pct:** The percentage of people who work at home (i.e. no transportation needed).
- **time_total:** The number of people who were sampled in the travel time to work survey.
- **time_0_29_pct:** The percentage of people who travel less than 30 minutes to work.
- **time_30_59_pct:** The percentage of people who travel between 30 and 59 minutes to work.
- **time_60_89_pct:** The percentage of people who travel between 60 and 89 minutes to work.
- **time_more_90_pct:** The percentage of people who travel more than 90 minutes to work.

1.2.4 Data import code and summary

Data has been imported from the codes in the previous sections.

```
export_summary_table(dfSummary(cbdata_tidy))
```