

#### Publicly Available Microdata:

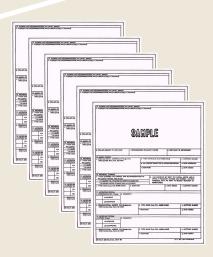
Why It's Useful and How to Use It in R

By Hayley Brown

Research Associate Center for Economic and Policy Research

March 2021

#### What is microdata?



#### Summary data vs micro data

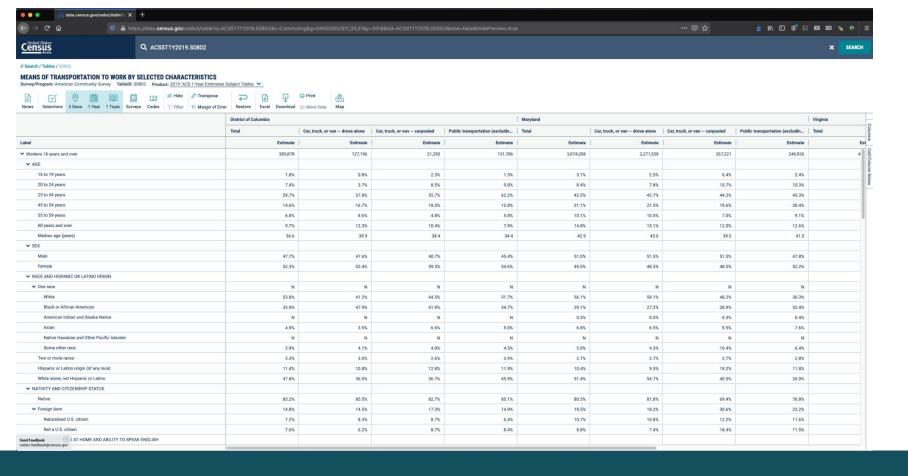
#### Aggregate data (summary data)

- Higher level data that results from grouping individual responses in a certain way
- Must balance data integrity with privacy protection
- May be available for smaller geographies, but at the expense of dimensionality
- Reduced dimensionality impedes certain types of analysis

#### Disaggregate data (*microdata*)

- Information at the level of individual responses
- Must balance data integrity with privacy protection
- Data may not be available for smaller subgroups or geographies
- Increased dimensionality and flexibility, well-suited to exploratory analysis

An example of summary data - <u>Table So802</u> from the Census website



#### MEANS OF TRANSPORTATION TO WORK BY SELECTED CHARACTERISTICS

Survey/Program: American Community Survey TableID: S0802 Product: 2019: ACS 1-Year Estimates Subject Tables 💌

	$\square$	0	YR			123			$\supseteq$	4	T.	Print	2
Notes	Selections	3 Geos	1 Year	1 Topic	Surveys	Codes	∀ Filter	★ Margin of Error	Restore	Excel	Download	More Data	Мар

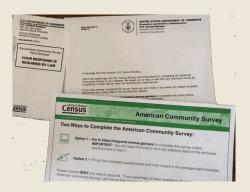
	Dist	District of Columbia				
	Tota	al	Car, truck, or van drove alone	Car, truck, or van carpooled	Public transportation (excludin	
Label		Estimate	Estimate	Estimate	Estimate	
➤ Workers 16 years and over		385,878	127,196	21,290	131,786	
<b>∨</b> AGE						
16 to 19 years		1.8%	0.8%	2.3%	1.5%	
20 to 24 years		7.4%	3.7%	8.5%	9.8%	
25 to 44 years		59.7%	57.8%	55.7%	62.2%	
45 to 54 years		14.6%	16.7%	18.3%	12.8%	
55 to 59 years		6.8%	8.6%	4.8%	5.8%	
60 years and over		9.7%	12.3%	10.4%	7.9%	
Median age (years)		36.6	39.9	38.4	34.4	

	District of Columbia	District of Columbia				
	Total	Car, truck, or van drove alone	Car, truck, or van carpooled	Public transportation (excludin		
Label	Estimate	Estimate	Estimate	Estimat		
➤ Workers 16 years and over	385,878	127,196	21,290	131,78		
> AGE						
> SEX						
➤ RACE AND HISPANIC OR LATINO ORIGIN						
✓ One race	N	N	N			
White	53.8%	41.2%	44.5%	51.7		
Black or African American	33.9%	47.9%	41.9%	34.7		
American Indian and Alaska Native	N	N	N			
Asian	4.9%	3.5%	6.6%	5.0		
Native Hawaiian and Other Pacific Islander	N	N	N			
Some other race	3.9%	4.1%	4.0%	4.5		
Two or more races	3.3%	3.0%	2.6%	3.9		
Hispanic or Latino origin (of any race)	11.4%	10.8%	12.8%	11.9		
White alone, not Hispanic or Latino	47.8%	36.5%	36.7%	45.9		

What if we want to know how many Black people age 50 and older take public transit to work?

We need to use microdata!

# Microdata from publicly available surveys



# Federally administered household surveys in the US

An excellent use of federal tax dollars

- Survey data is different from administrative data
- Substantial variation in scope, survey type, sample size, administration (frequency, means of contact)
- Typically include both household and person level items

#### Some examples of US household surveys

- Current Population Survey (CPS): Primary source of US labor force statistics; basic survey conducted monthly, supplemental surveys administered less frequently
- American Community Survey (ACS): Annual survey that replaced the long-form census, largest household survey administered by the US Census Bureau
- National Health Interview Survey (NHIS): Annual survey encompassing a wide range of health status and health care utilization measures
- American Time Use Survey (ATUS): Annual survey measuring how much time respondents spend on various activities

#### A few more examples of US household surveys

- Household Pulse Survey (HPS): New longitudinal survey designed to measure socioeconomic impact of the COVID-19 pandemic; administered online
- Survey of Income and Program Participation (SIPP): Continuous national panel survey with detailed information on household income dynamics and participation in government transfer programs
- American Housing Survey (AHS): Largest regular national housing sample survey in the US, conducted every other year
- ...and <u>many more!</u>

#### Considerations when deciding which survey to use

- **Population of interest:** Geography, age, labor force status, housing type, etc.
- Available variables: Does the survey contain the variables I need to perform my analysis?
- Sample size: Larger samples needed to study smaller subgroups and geographies, and to perform certain types of statistical analysis (e.g. matching)
- Survey type: How important is longitudinal data to your analysis?
- Availability and ease of use: Is there a cleaned up version available or will you need to clean the raw data yourself?

# Accessing Public-Use Microdata for Household Surveys



#### Raw data vs cleaned extracts

#### Raw public-use microdata

- Tends to require a lot of cleaning, especially to harmonize variables that undergo changes between sample years
- Some datasets are not available any other way
- The data dictionary is your best friend

#### Cleaned extracts

- Raw government data has been processed, cleaned, and harmonized by a third party
- May contain additional variables based on pre-existing ones; important to be aware of edits made to underlying data
- Typically much easier to use

#### Accessing raw public-use microdata

- Agency websites: You can typically find the public-use microdata online via the agency or agencies that administer/ sponsor the survey
- Limited API availability for now
- For some datasets and sample years, increased <u>customization</u> for downloads
  - This is especially useful for larger datasets like ACS
  - Some data is still only available via FTP, which means you have to download the whole thing before you can work with it

### Some sources of US household survey microdata extracts

- **IPUMS:** Large-scale project that allows users to download customized extracts; incredibly comprehensive, with many additional useful variables, including geographic crosswalks. They also have their own R package, ipumsr.
- **CEPRdata:** Ongoing project of CEPR, limited number of datasets but working on improvements; program files used to create extracts are available but are in Stata
- **EPI Microdata:** Some variables similar to CEPRdata, has CPS Basic and ORG data updated regularly

#### Using Survey Microdata in R





### Revisiting this line of questioning:

What share of people age 50 and older who take public transit to work are Black? What share are Hispanic?

- You can follow along on GitHub:
   <a href="https://github.com/hcbrown/TR\_ACSexercise">https://github.com/hcbrown/TR\_ACSexercise</a>
- We will be working with a subset of the CEPRdata ACS PUMS extract.
   Specifically, we will be using the 2019 1-year sample for DC, MD, and VA.
   You can download the data here:
   <a href="https://ceprdata.org/wp-content/acs/data/cepr\_acs\_2019\_dmv.csv.zip">https://ceprdata.org/wp-content/acs/data/cepr\_acs\_2019\_dmv.csv.zip</a>
- This exercise uses RStudio and the dplyr and pollster packages

#### • Importing the data:

- Setting directories:
  setwd("/your/path/to/your/data/")
- Reading without decompressing will be important when working with larger files:

```
acs19dmv <- read.csv(unz("cepr_acs_2019_dmv.csv.zip", "cepr_acs_2019_dmv.csv"),
header = TRUE, sep = ",")</pre>
```

#### Selecting only the variables we care about:

- We need age, race/ethnicity, whether the person uses public transit to commute to work, and the person weight.
- Note that two of these four variables are CEPRextract creations, rather than ACS originals.

```
acs19dmv_sub <- acs19dmv %>%
  select(age,
     wbhao,
     pubtran,
     perwgt)
```

- Create a new variable combining age and race/ethnicity:
  - Use mutate to create the new variable
    - We've restricted our new variable, agerace, to those in-universe for pubtran

```
acs19dmv_sub <- acs19dmv_sub %>%

mutate(agerace = case_when(age >= 50 & wbhao == 'White' & (pubtran == 0 | pubtran == 1) ~ 'White, age 50+',

age >= 50 & wbhao == 'Black' & (pubtran == 0 | pubtran == 1) ~ 'Black, age 50+',

age >= 50 & wbhao == 'Hispanic' & (pubtran == 0 | pubtran == 1) ~ 'Hispanic, age 50+',

age >= 50 & wbhao == 'AAPI' & (pubtran == 0 | pubtran == 1) ~ 'AAPI, age 50+',

age >= 50 & wbhao == 'Other' & (pubtran == 0 | pubtran == 1) ~ 'Other, age 50+'))
```

#### • Running a crosstab:

 Eliminate the rest of the observations with NA values for agerace, the variable we just defined

```
acs19dmv_sub <- acs19dmv_sub[!is.na(acs19dmv_sub$agerace),]</pre>
```

Run crosstab for agerace and pubtran

```
shares <- crosstab(acs19dmv_sub, pubtran, agerace, w = perwgt)</pre>
```

 Note the use of perwgt. It is important to use the appropriate weights when tabulating with microdata.

## Our answered question

Pre-pandemic in the District of Columbia, Maryland, and Virginia: Black workers age 50 and older were more likely to commute to work via public transit than White workers age 50 and older.

#### Workers Age 50 and Older in DC, MD, and VA, by Race/Ethnicity and Use of Public Transit to Commute to Work

	Does not use public transit to commute to work	Uses public transit to commute to work	All
White, age 50+	64.9%	38.8%	63.3%
Black, age 50+	21.5%	44.7%	22.9%
Hispanic, age 50+	6.4%	8.6%	6.5%
AAPI, age 50+	6.6%	7.2%	6.7%
Other, age 50+	0.6%	0.7%	0.6%
Total	100%	100%	100%