00 Sustainability Trends Among Disadvantaged Communities

Exploring the Climate and Economic Justice Screening Tool (CEJST) Data

Stefani Langehennig Zach del Rosario

Before starting the activity, please use this QR code to access a survey. Even if you choose not to participate in the associated study, you are still welcome to do the activity!



Please do not move on until your instructor tells you to.

Overview

As climate change continues to impact the world in which we live, numerous initiatives have been started to better understand the influence it has on individuals and communities. One of those initiatives stems directly from an Executive Order (EO) issued by President Joe Biden in January 2021. The EO resulted in the Council on Environmental Quality creating a tool by which the public can track various burdens across a number of communities. The primary aim of the tool is identify and subsequently help communities disadvantaged by these burdens in government social programs.

The Climate and Economic Justice Screening Tool (CEJST) is the result of the EO. While the tool established by the Council on Environmental Quality covers a number of burdens (health, transportation, and workforce development, for example), this activity will focus on the sustainability aspects of the tool, including climate change, energy, and legacy pollution burdens on communities.

To set the stage for this activity, we are going to use the CEJST data to explore whether there is a relationship between the energy burden percentile in census tracts and the share of population of Blacks or African-Americans in census tracts. Below, we will explore the dataset, as well as our variables of interest, in-depth.

Dataset

The data used for this analysis comes from the CEJST website. The columns (variables) we are most interested in for better understanding these data are:

- The *energy burden percentile*, which captures the percentile of energy cost as well as energy-related pollution within a census tract.
- The percent of African-American or Black alone, which captures the percent of African-American or Black individuals in a census tract.

First, we will load our data and clean up the variable names using the various packages available to us in the tidyverse.

```
# Import tidyverse
library(tidyverse)

# Load CEJST data and create a new dataframe called 'df_raw'
filename <- "../data/1.0-communities.csv"
df_raw <- read_csv(filename)</pre>
```

```
# Create a new dataframe called 'df_data' with new column names
  df data <-
    df_raw %>%
    janitor::clean_names()
  # Select only those columns we'll use in this activity
  df_data %>%
    select(
      census_tract_2010_id,
      percent_black_or_african_american_alone,
      energy_burden_percentile
    ) %>%
    head(3)
# A tibble: 3 x 3
  census_tract_2010_id percent_black_or_african_america~1 energy_burden_percen~2
  <chr>>
                                                     <dbl>
                                                                             <dbl>
1 01001020100
                                                      0.07
                                                                                49
```

We will focus on a few columns in this dataset:

2: energy_burden_percentile

2 01001020200

3 01001020300

• census_tract_2010_id: Each row in this dataset corresponds to a *census tract*; this is a small geographic region of the U.S. chosen to represent a consistent number of persons. Census tracts contain about 4000 people, though may contain as few as 1200 and as many as 8000 people.

i abbreviated names: 1: percent_black_or_african_american_alone,

0.57

0.24

6

68

- percent_black_or_african_american_alone: This reports the percent of people in the census tract who are Black or African American.
- energy_burden_percentile: This reports the percentile of energy burden for each census tract. Energy burden is computed as the average annual cost of energy divided by the average household income within the census tract—this is a measure of how "burdened" a household is by energy expenses. A larger energy burden means a household needs to spend more of its income on energy bills.
 - The percentile is then computed as the ordering of energy burden values for each census tract: The 0th percentile corresponds to the smallest value of energy burden, while the 100th percentile corresponds to the largest value of energy burden.

Next, let's conduct EDA to better understand our variables of interest, energy_burden_percentile and percent_black_or_african_american_alone.

Exploratory Data Analysis (EDA)

To begin, let's subset our data so we can get our descriptive statistics for our variables of interest.

```
# Subset full dataframe
df_small <-
    df_data %>%
    select(energy_burden_percentile, percent_black_or_african_american_alone)

# Take a look at the subset of data
glimpse(df_small)

Rows: 74,134
Columns: 2
```

\$ percent_black_or_african_american_alone <dbl> 0.07, 0.57, 0.24, 0.05, 0.18, ~

<dbl> 49, 6, 68, 63, 38, 57, 72, 46,~

Let's use this subset to get our measures of spread and central tendency.

```
summary(df_small)
```

\$ energy_burden_percentile

```
energy_burden_percentile percent_black_or_african_american_alone
Min.
       : 0.00
                          Min.
                                  :0.0000
                          1st Qu.:0.0100
1st Qu.: 24.00
Median: 49.00
                          Median : 0.0400
Mean
       : 49.55
                          Mean
                                  :0.1341
3rd Qu.: 75.00
                          3rd Qu.:0.1500
       :100.00
Max.
                          Max.
                                  :1.0000
NA's
       :1054
                          NA's
                                  :745
```

Synthesizing Descriptive Statistics

What are some take-aways from our descriptive statistics for our variables of interest? Anything concerning or interesting?

Next, we can plot the percent of Black or African-Americans against the energy burden percentile to see if there is a negative or positive trend between the two variables.

As the figure shows, there is a positive relationship between the percent of Black or African-Americans living in a census tract and the energy burden percentile in the respective census tract.

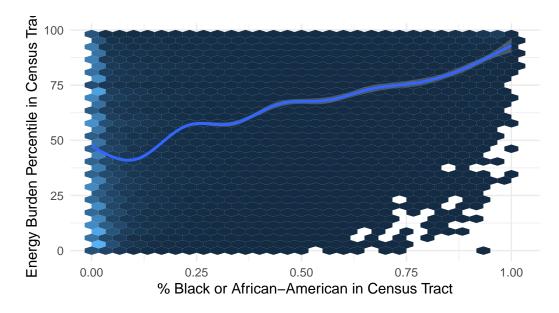


Figure 1: Basic scatterplot of the CEJST data.

Visualizing Variables of Interest

What other ways might we choose to visualize our variables of interest? Is there anything else concerning or interesting based on our visualizations?

Next Steps

We now have a better understanding of sustainability trends among disadvantaged communities using just a few of the variables in the CEJST dataset. Specifically, we explored the relationship between the percent of Black or African-Americans living in a census tract and the energy burden percentile in the census tract. Our high-level exploratory data analysis uncovered a positive relationship: Black or African-Americans that live in a census tract appear to experience a higher energy burden.

Other Trends

How might the trend differ by state? Why do you think they will or will not differ from the overall trend observed above?

With this information in hand, we will use these data, as well as other variables in the CEJST dataset, to formulate hypotheses. We will test our hypotheses in both a frequentist and Bayesian framework, comparing the application of both approaches across the different stages of our analysis with the end goal being general inference.

Coming up next: When the instructor tells you, please start working through 01 Introduction to Different Statistical Paradigms with the other students at your table. Please do not move on until your instructor tells you to.