



Homeownership & Poverty in the United States

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Introduction

- We knew that we wanted to investigate the rising rates of poverty in America
- The main point that we wanted to investigate was the impact that Home Ownership had on poverty
- We felt that providing easier ways to obtain housing would lead to lower poverty rates across the country



Introduction (contd.)

- The main statistic that we used was correlation
- We looked for a strong negative correlation
- We investigated data across 5 different races and a total value
- We selected data over the years 2005-2017



Data Preparation

- We were able to find 2 data sets that allowed for us to run our programs
 - [Diversitydatakids.org](https://diversitydatakids.org)
- Poverty & Homeownership by state yielded too few data points for us to make any meaningful conclusions
- These same rates sorted by county gave us over 7,000 data points once we cut out the empty areas



Data Preparation

geoid	name	year	total_est	aian_est	api_est	asian_est	black_est	hisp_est	nhisp_est	nhopi_est
05000US01001	Autauga County, Alabama	2005-2009	10.306	7.253	0	0	27.565	12.747	10.255	None

api_est	numeric	Estimate; Asian or Pacific Islander Alone
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Data Preparation

- Once we exported both of the datasets as csv files we were able to combine them into one large, master database that allowed us to run our calculations more smoothly
- From here, we were able to clean it to get more accurate conclusions



Data Cleaning & Pre-Processing:


- Data cleaning was short
- Main concern was to remove blank data cells
- Used Excel's filter options to remove all the blank cells
 - This also helped our csv file become smaller so Python could run the necessary calculations
- Other main issue was deciding how many races to use
 - Started with 17 different columns, shortened to six
- This left a much more easily digestible data set



Data Analysis

- To compute the necessary correlation coefficient (r-value) we first converted the downloaded Excel file into a CSV file
- We then designed a Python Program Called Correlation Calculator that reads the CSV file and returns an r-value
- It would gather all the values of the Homeownership column, treat them as X values, and compare them to the corresponding Poverty rate (which represented Y values)
 - We made use of a function that could repeat the results whenever a new pair of columns were input meaning this program could easily be retooled to seek out correlations with other groups
- Figure 1 is the results

Figure 1



Correlation Coefficients	
Group	r-value
Total	-0.4244
African American	-0.2250
Hispanic	-0.2719
Asian and Pacific Islander	-0.3105
Non-Hispanic white	-0.3509
White	-0.3507



Data Analysis (cont)

- In order to show a strong negative correlation values should be less than $-.75$ but an r-value of $-.5$ would be significant
- As Figure 1 showed, while each of our variables had negative correlations, none of them reached a threshold of significance
- Our best correlation was in the “totals” column at $-.4244$ while the group that had the lowest correlation was African Americans at $-.225$



Data visualization

- We created scatter plots with Homeownership rates on the X axis and Poverty rates on the Y axis
- Figure 3 indicates that utilizing all the data points would render our models as one big blob
- Because creating scatterplots with a little under 8000 data points would be difficult to read, we instead decided to randomly select 800 rows of data and use Excel's built-in visualization tools to generate the scatter plots

Example of all data points

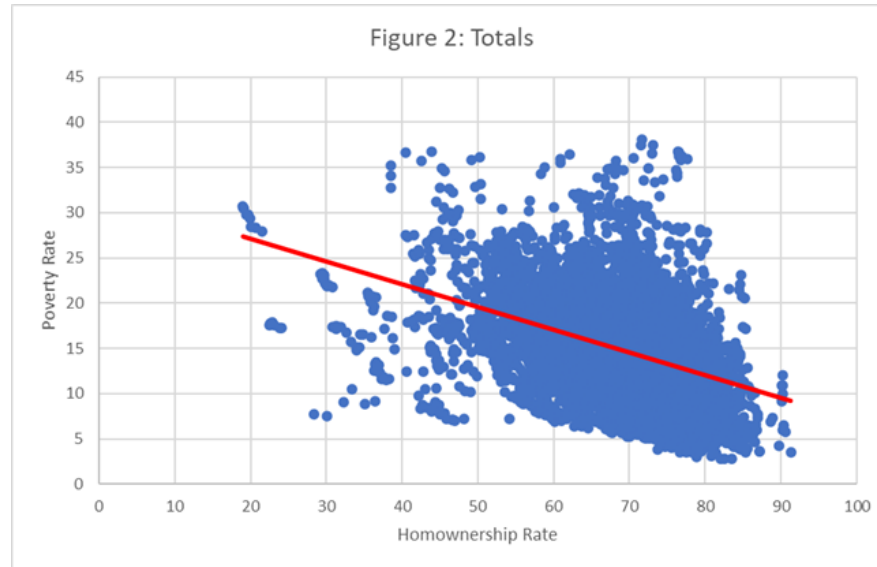
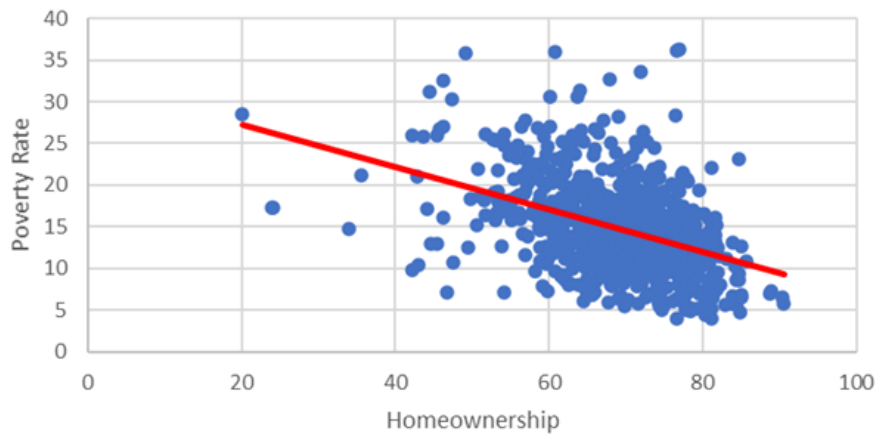


Figure 2 showcases how all the data points are laid out on a scatter plot. Each blue dot represents a data point from a specific county for a specific period. The x-axis (the independent variable) is the homeownership rate while the y axis (dependent variable) is the poverty rate. Unfortunately the mass of points makes it difficult to read

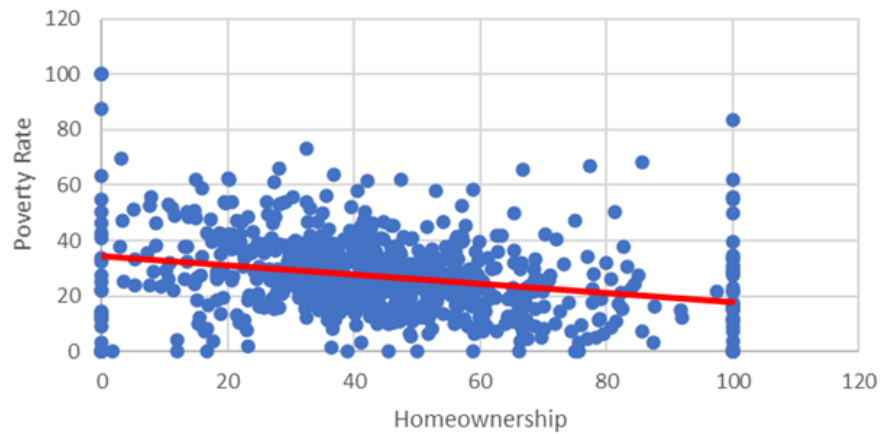


Figure 3: Total



Sample size of 800 for the “Total” columns

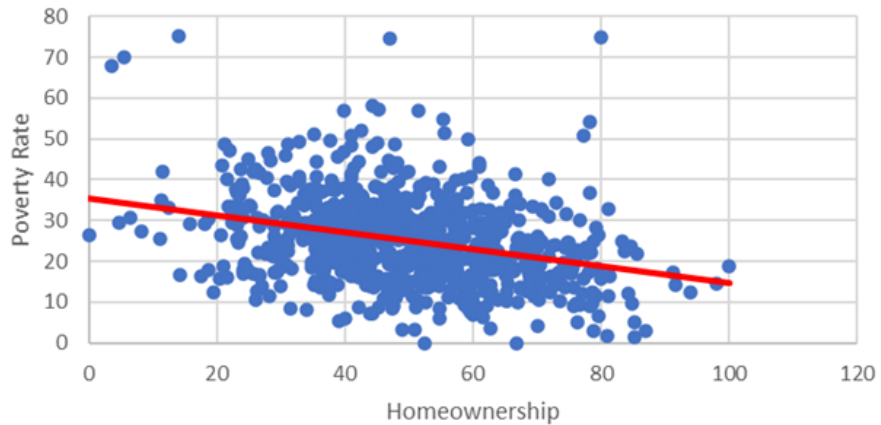
Figure 4: African American



Sample for the “African American” columns

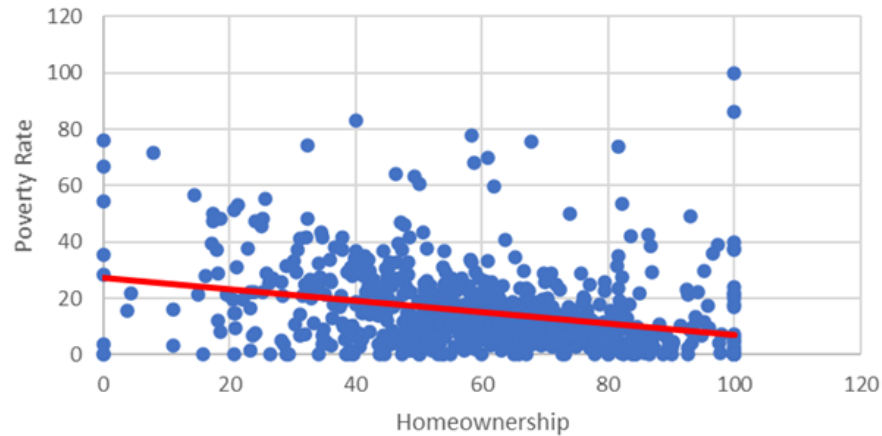


Figure 5: Hispanic



Sample for the “Hispanic” columns

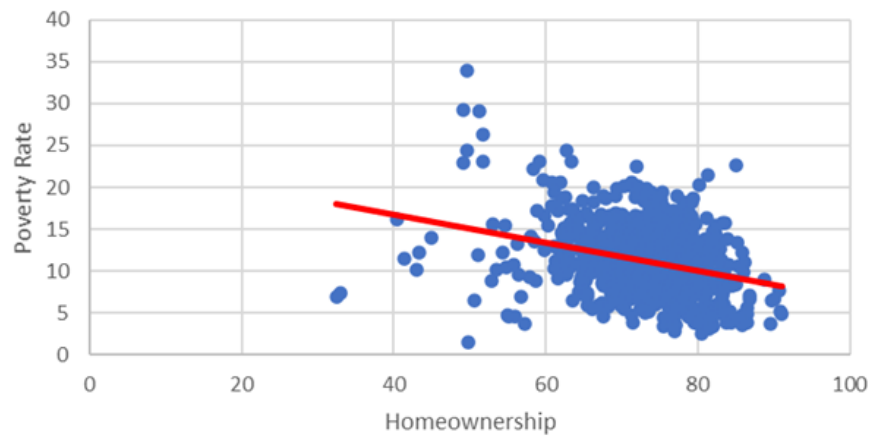
Figure 6: Asian & Pacific Island



Sample for the “Asian & Pacific Island” columns

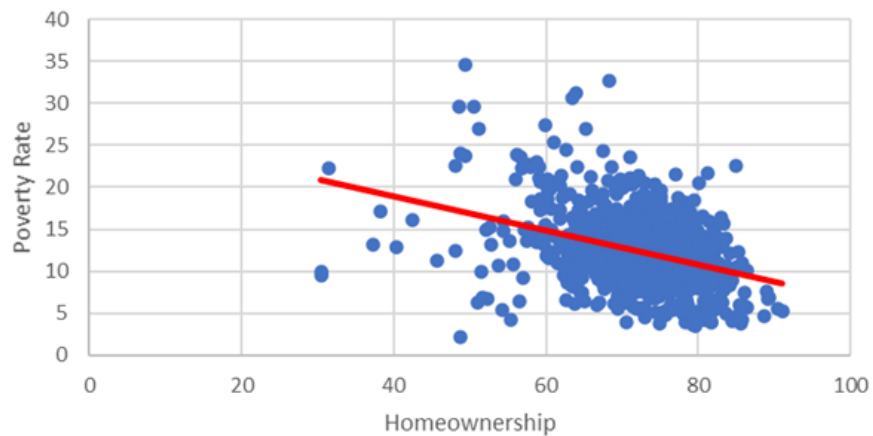


Figure 7: NonHispanic White



Sample for the “NonHispanic White” columns

Figure 8: White



Sample for the “White” columns



Data Visualization (Cont)

- The scatter plots show a weak linear relationship between our two variables
- Very few data points touch the trendline, and outliers are common
- All trend lines are negative, showing that while the correlation coefficient might not be strong, it is still negative



Conclusion

- Poverty is a huge issue in the United States
- We believed there may be correlation between homeownership and poverty
- We conducted research and implemented programs to test our hypothesis
- We found negative correlation, but not enough to completely justify our claim
- Further research should be conducted as homeownership alone is not significant enough