



Identifying Counties for a Food Access Program



Using Public Data on Health Outcomes,
Socioeconomic and Sociodemographic Statuses,
and Access to Healthy Food



Introduction

N1 Health's client, a national Medicare Advantage plan, is seeking to understand how best to use its resources to address food access challenges in its membership.

1. Where should we deploy a food access program?
1. Where should we deploy a food access program?
2. Which segment of the population might benefit the most from the program?
3. What do we know about the potential impact of the food program on medical costs or utilization?

Data Wrangling

Pivoting raw tables

Preliminary feature selection using `df_fda_vars`

- Using Excel

- Why use this?

- Why those variables?

Data Slicing

- Using `col_feat` to create `df_fda_full`

Exploratory Data Analysis

Feature Engineering

- Working only with percent change instead of PCT_POP
 - Dim Reduction
 - Consistent data types
 - Reflects momentum in the data
- Creation of Need Ranking
- Transformation

Variables in the Need Ranking

1. PERPOV10
2. CHILDPOVRATE15
3. PERCHLDPOV10
4. PCT_BIPOC_10
5. PCH_FOODINSEC_12_17
6. PCT_LACCESS_BIPOC15

Model Training/Deployment

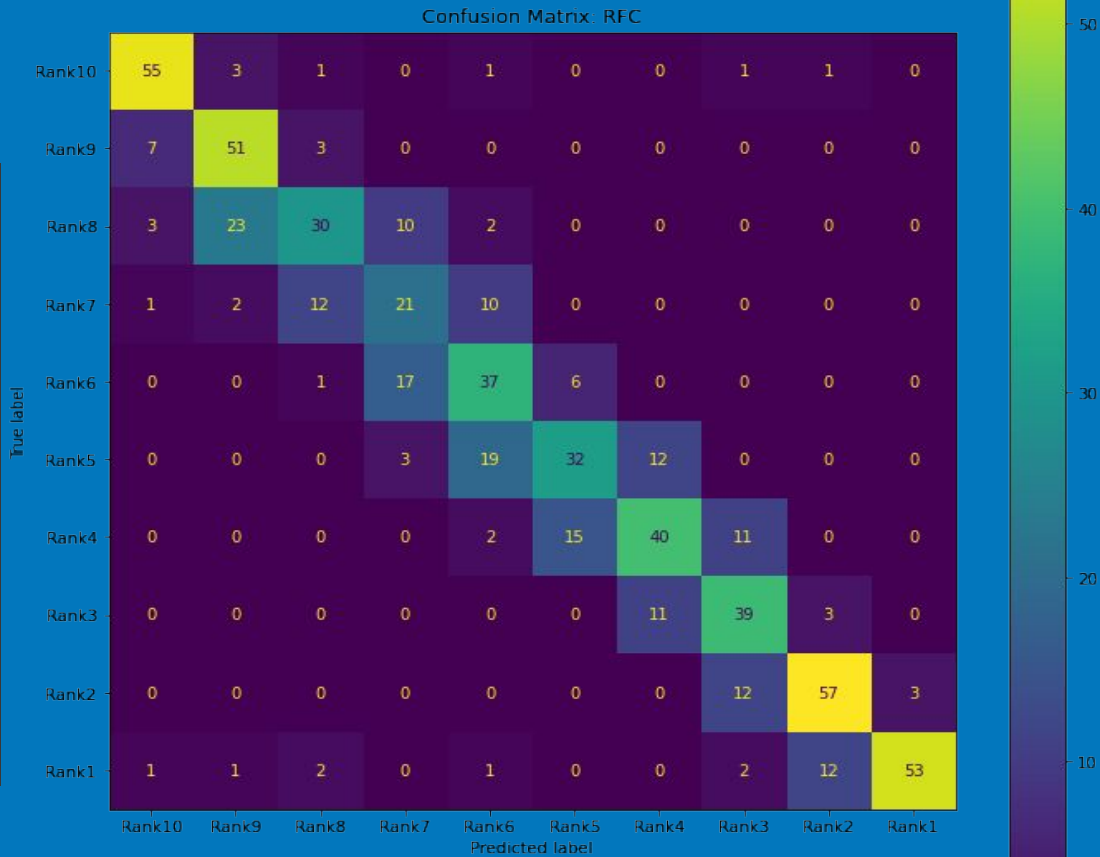
RFC

Randomized Search with Cross Validation

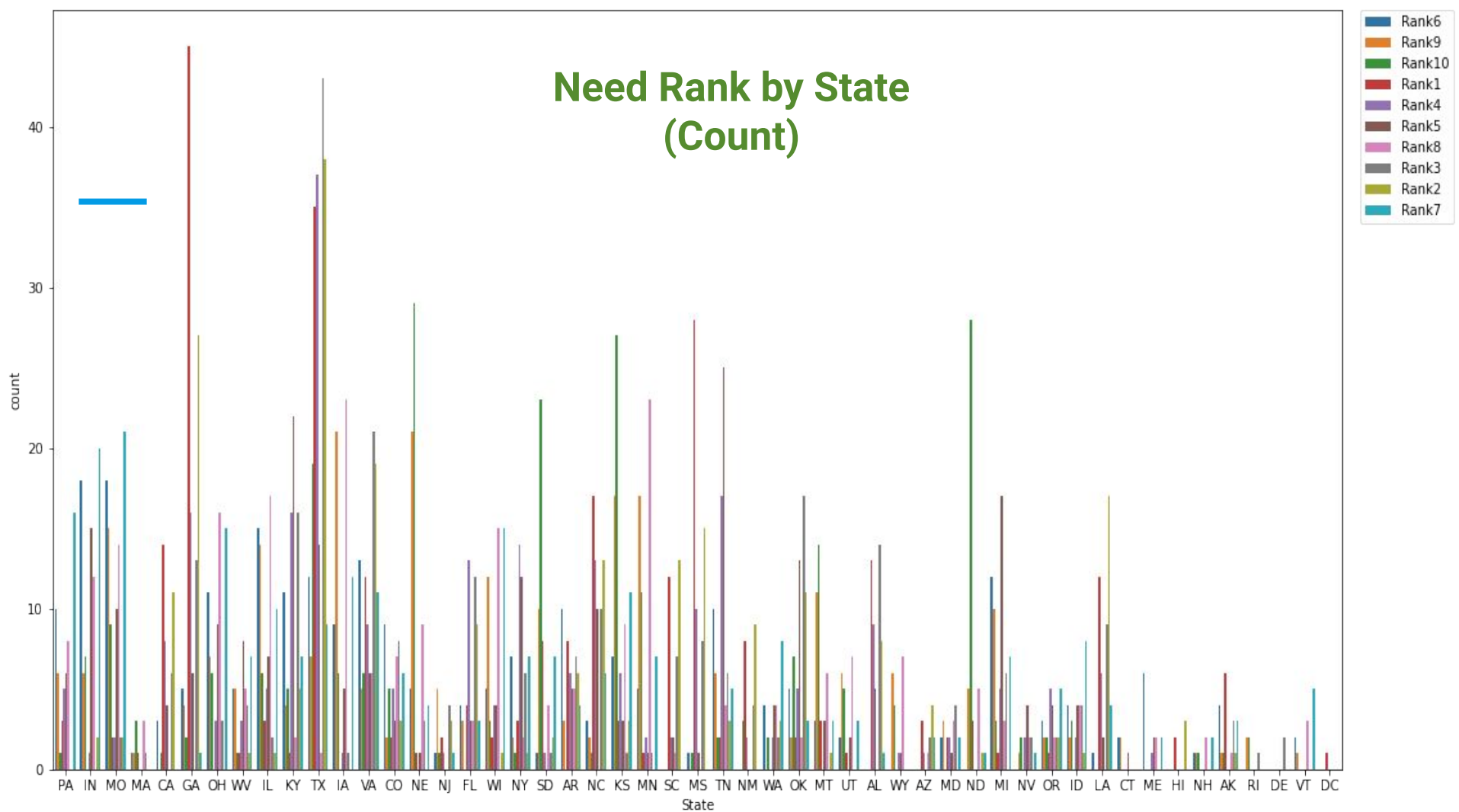
Exploring the Results

Random Forest Classifier

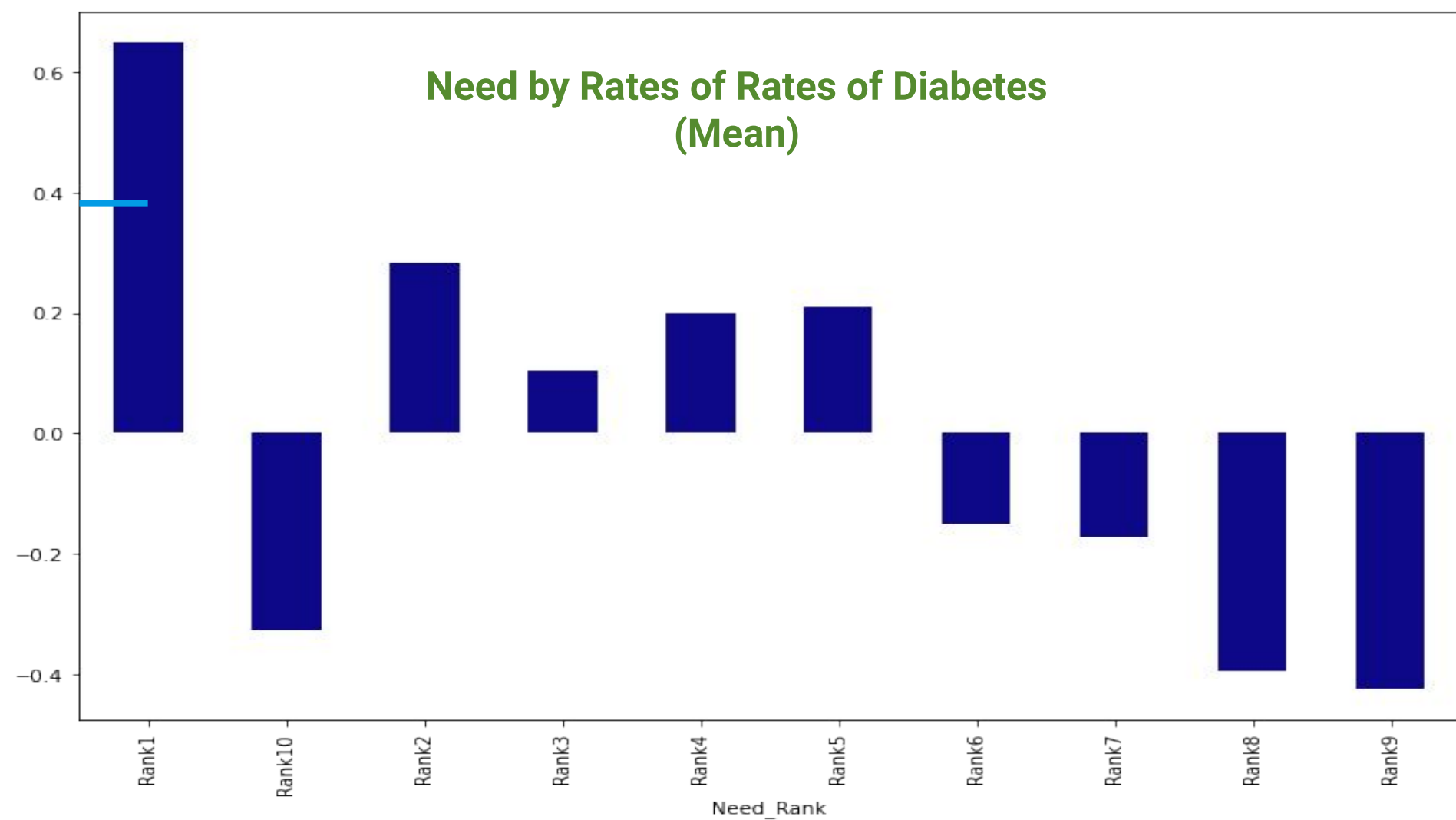
| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| Rank10 | 0.95 | 0.74 | 0.83 | 72 |
| Rank9 | 0.82 | 0.89 | 0.85 | 62 |
| Rank8 | 0.78 | 0.79 | 0.79 | 72 |
| Rank7 | 0.60 | 0.74 | 0.66 | 53 |
| Rank6 | 0.63 | 0.59 | 0.61 | 68 |
| Rank5 | 0.60 | 0.48 | 0.54 | 66 |
| Rank4 | 0.51 | 0.61 | 0.56 | 61 |
| Rank3 | 0.41 | 0.46 | 0.43 | 46 |
| Rank2 | 0.61 | 0.44 | 0.51 | 68 |
| Rank1 | 0.64 | 0.84 | 0.72 | 61 |
| accuracy | | | 0.66 | 629 |
| macro avg | 0.66 | 0.66 | 0.65 | 629 |
| weighted avg | 0.67 | 0.66 | 0.66 | 629 |



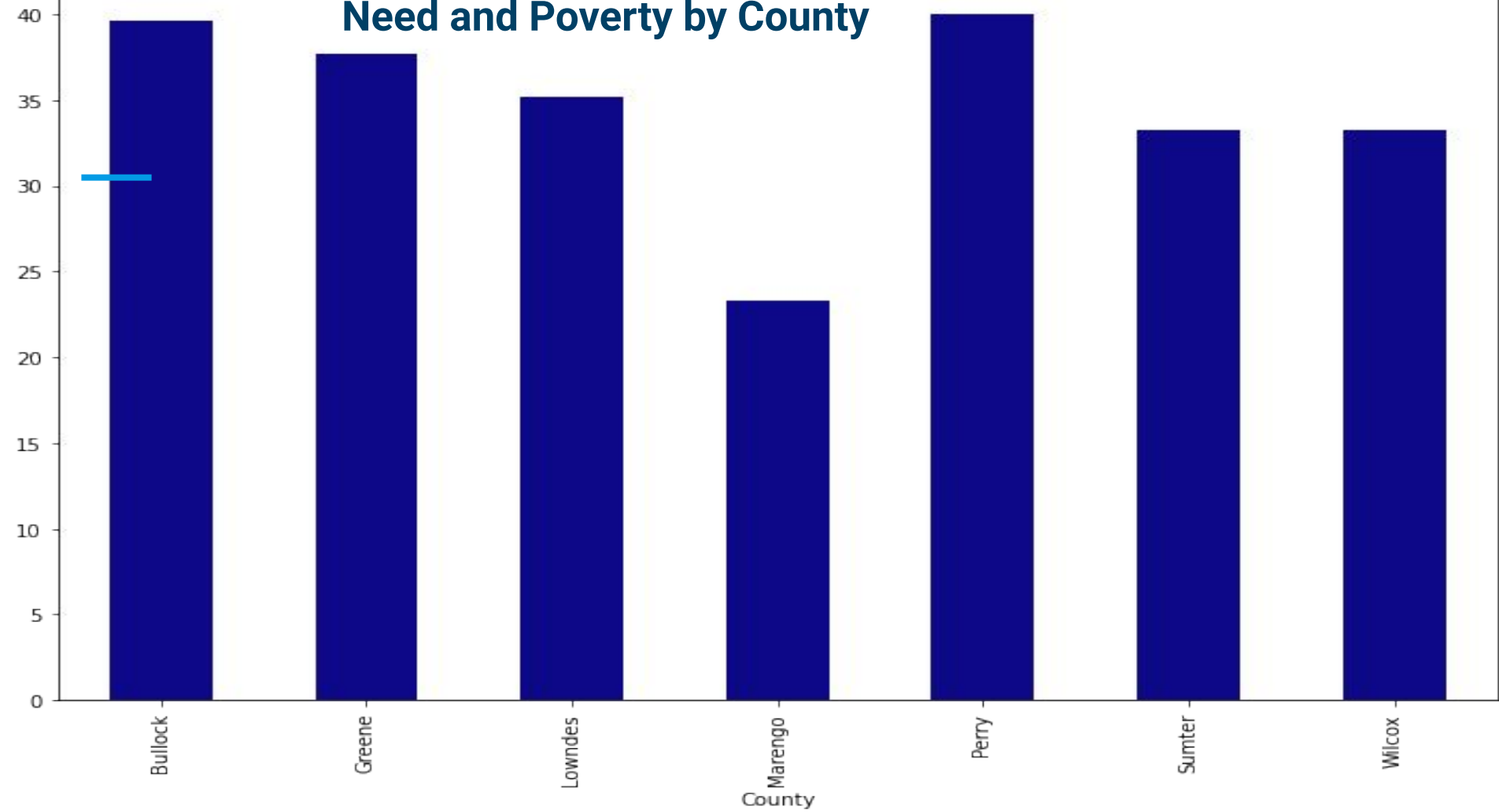
Need Rank by State (Count)



Need by Rates of Rates of Diabetes (Mean)



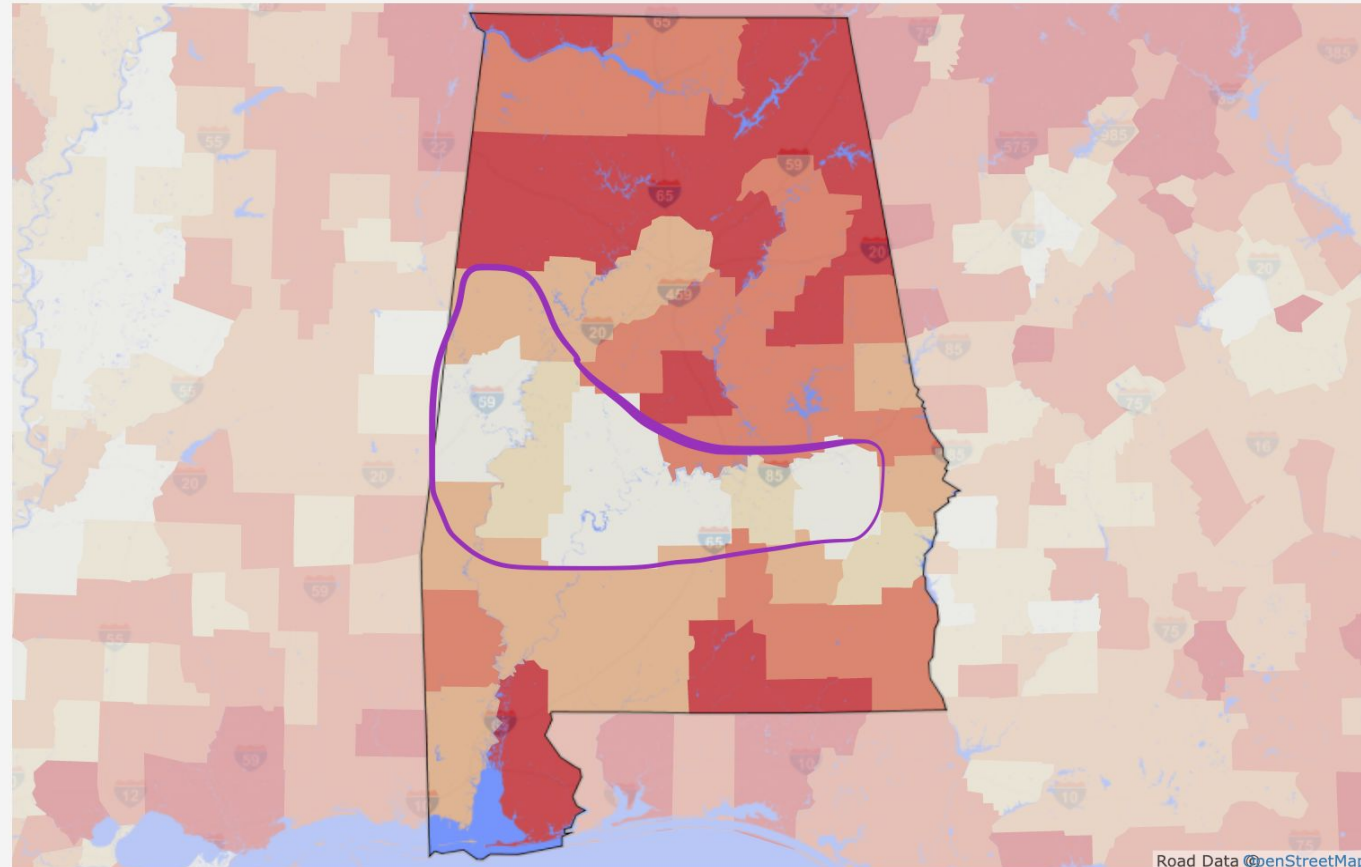
Need and Poverty by County



White Race and Ethnicity by County

#7

Whites (non-Hispanic) as a percentage of the population (%):



Limitations/ Future Work

- Health outcome data (CDC)
- Ranking System
- Machine Learning to Predict Food Insecurity by County
- Integration with CMS data
 - ML to predict expenditures
- What Sort of Program?
 - New or Existing Program?
- Communities Impacted:
 - BIPOC
 - Experiencing Poverty
 - Highest Rates of Metabolic Syndromes