

Predicting Successful American Counties

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Project Goals:

- ▶ Predict the successful American Counties
- ▶ Understand what variables makes some Counties more successful
- ▶ Create a useful model for forecasting
- ▶ Get a high adj. R squared!

Design and Approach

Define “Successful” - Combine population growth with gini coefficient

- ▶ Measure desirability of an area and its inequality
- ▶ “Success” = $0.5norm_populationgrowth + 0.5norm_gini$

Population vs. Gini

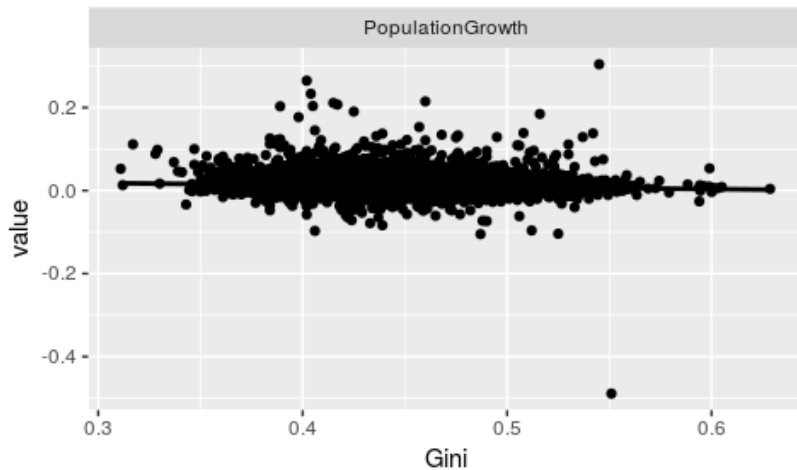


Figure 1: Population Growth

Used data from Census bureau

- ▶ Unemployment Rate
- ▶ Median Income
- ▶ Poverty Rate
- ▶ Housing Costs
- ▶ Mortgage Costs
- ▶ Bachelor Rate
- ▶ Population
- ▶ Fertility

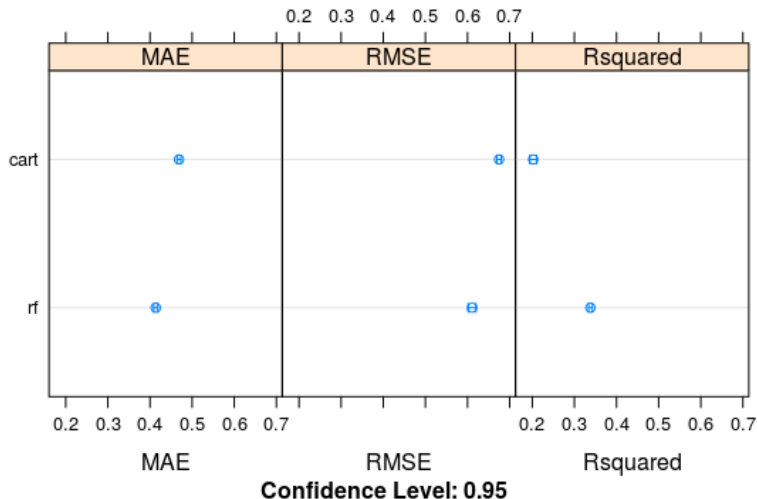
Methods

We used three main algorithms:

- ▶ Linear Regression
- ▶ Regression Tree
- ▶ Random Forest

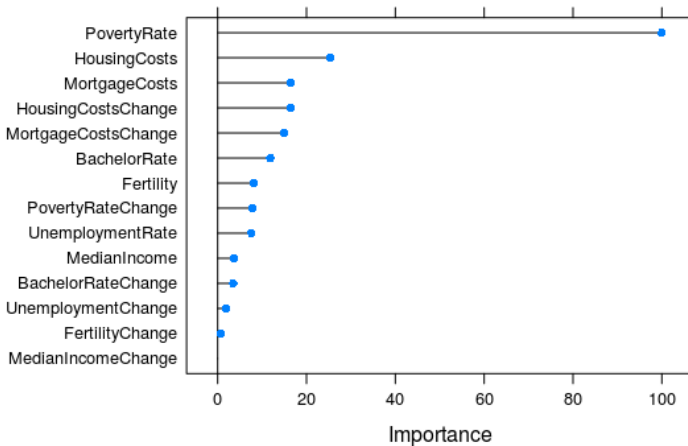
Methods cont'd

- ▶ Of the three, Random Forest performed best.
- ▶ However, it still had an adjusted R-squared of ~33.78%



Results

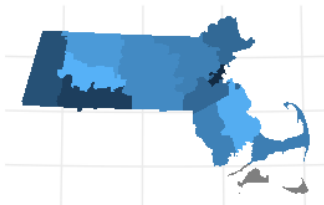
- ▶ The Poverty rate was (by far) the strongest predictor for “Success”
- ▶ Mortgage and Housing Costs were the two next most significant



Results Graphically

- Our predicted was not (too) far from the actual.

Real Success in Massachusetts



Predicted Success in Massachusetts

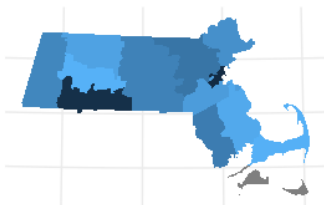
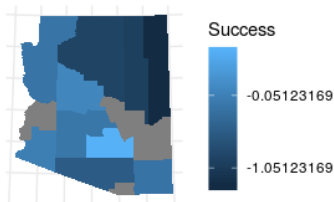


Figure 2: Massachusetts

Results cont'd

- Note that there were some missing observations, so some counties are gray

Real Success in Arizona



Predicted Success in Arizona

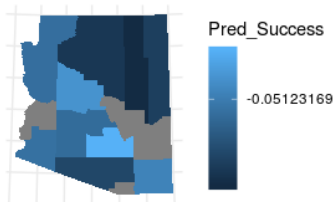


Figure 3: Arizona

Conclusions

- ▶ Our model explains very little of the deviation in our output.
- ▶ However, our finding that the Poverty Rate has an outside effect on our output is significant.
- ▶ We can further explore other variables that could be correlated to “Success” (weather? crime?)
- ▶ We could also explore how the variables we studied interact with the Poverty rate.
- ▶ We could also forecast the variables we have and predict the “future” success of American counties.