Business Types and Poverty Rates

STAT 320

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Poverty / Intro

- The poverty rate in the United States as of 2021 is 11.6%
- Over 38 million people cannot afford the basic needs for their lives

- Goal:
- To examine the various factors that contribute to poverty in specific counties in the United States.



"National Poverty in America Awareness Month: January 2023." Census.Gov, Jan. 2023,

www.census.gov/newsroom/stories/poverty-awareness-month.ht ml.

Davies, Dave. "Private Opulence, Public Squalor: How the U.S. Helps the Rich and Hurts the Poor." NPR, 21 Mar. 2023, www.npr.org/sections/health-shots/2023/03/21/1164275807/pove rty-by-america-matthew-desmond-inequality.

Our Question?

How is poverty rate impacted by the rate of different business types?

We hypothesize that some types of business, such as healthcare will increase the wealth of a county, leading to a lower poverty rate for that county, while other types of business like vice related businesses will lead to a higher rate of poverty for that county.

The Data

- Environmental Protection Agency Data from 2013
- All counties in the USA
 - Dropped NA's
 - Split testing/training
- Used 9 business rates
 - Number of business of that type within the county divided by population

The Data

Quantitative Business types predictors:

- Vice
- Entertainment
- Education
- Negative food
- Positive food
- Healthcare
- Recreation
- Transportation
- Civic

Quantitative response:

Poverty Rate

Creating our Model

- Multiple linear regression
- Applied log transformation on the predictors and the results to fix normality
- Backward elimination
 - Dropped Vice because it was not significant

Model Results

- P-value < 2.2e-16
- R² on training = 20.39%
 - Testing = 19.2%
- All predictors very significant

Call:

```
lm(formula = trans_pov ~ trans_civic + trans_food_pos + trans_food_neg +
    trans_hc + trans_ed + trans_trans + trans_rec + trans_ent,
    data = training)
```

Residuals:

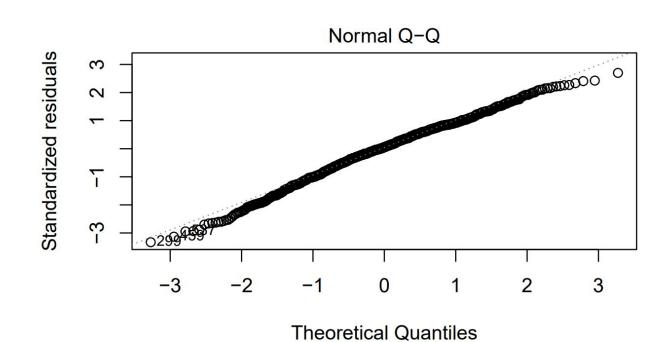
```
Min 1Q Median 3Q Max
-1.32859 -0.24725 0.02366 0.28140 1.07783
```

Coefficients:

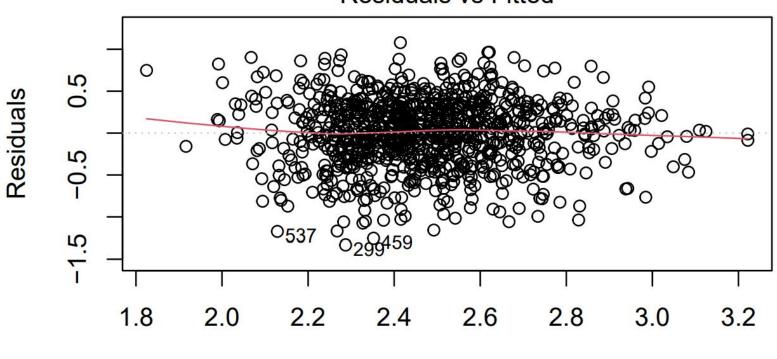
```
Estimate Std. Error t value Pr(>|t|)
(Intercept)
               3.82046
                          0.43177
                                   8.848 < 2e-16 ***
trans civic
               0.08712
                          0.02263 3.850 0.000126 ***
trans_food_pos 0.49437
                          0.06181 7.999 3.73e-15 ***
trans food neg 0.27095
                          0.05069 5.345 1.14e-07 ***
              -0.14033
                          0.04780 -2.936 0.003412 **
trans hc
              -0.07976
                          0.02800
                                  -2.848 0.004491 **
trans ed
              -0.06292
                          0.02283 -2.756 0.005970 **
trans_trans
trans_rec
              -0.09396
                          0.02713
                                  -3.463 0.000559 ***
              -0.22176
                          0.02998 -7.398 3.09e-13 ***
trans_ent
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Residual standard error: 0.4012 on 930 degrees of freedom Multiple R-squared: 0.2039, Adjusted R-squared: 0.1971 F-statistic: 29.77 on 8 and 930 DF, p-value: < 2.2e-16

Conditions



Residuals vs Fitted



Fitted values

Our findings

Higher Poverty

- Civic
- Positive food = 0.49
- Negative food

Lower Poverty

- Education
- Healthcare
- Transportation
- Recreation
- Entertainment = -0.22

Limitations and Conclusions

- We kept 1878 counties.
 - The dropped counties may have something in common
- Not applicable outside of the US
- Might change over time, might only work for 2013.
- More categories of business types
 - Be more specific with results
- Use case: More informed economic development strategies