# It Takes a Village

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A Presentation for the Coalition for Unprecedented Transformative Education for California Teachers and Students, 2024 Meeting

## Problem Statement

- Dropping out of high school impairs a child's opportunities for the rest of their life.
- You all want to help!
- We used government data to predict dropout rates from social conditions at the county level in California.
- ► Goals: Plan community actions, set political goals

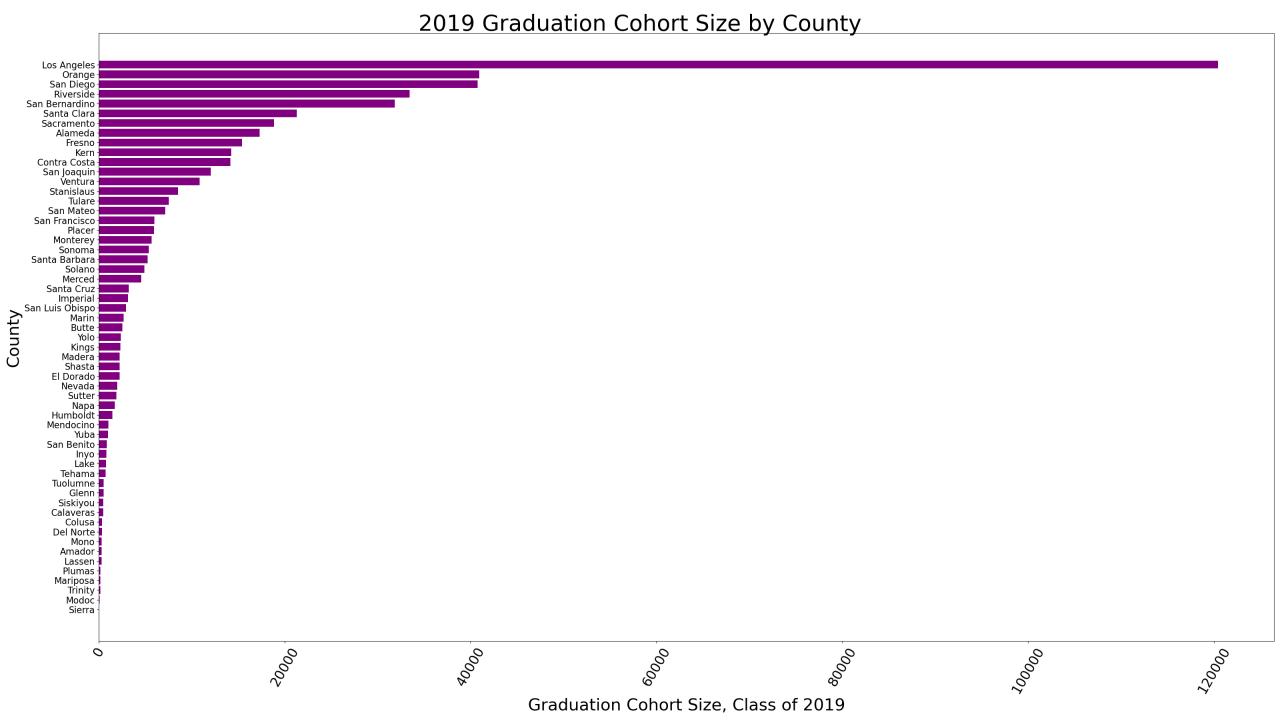
## Definitions

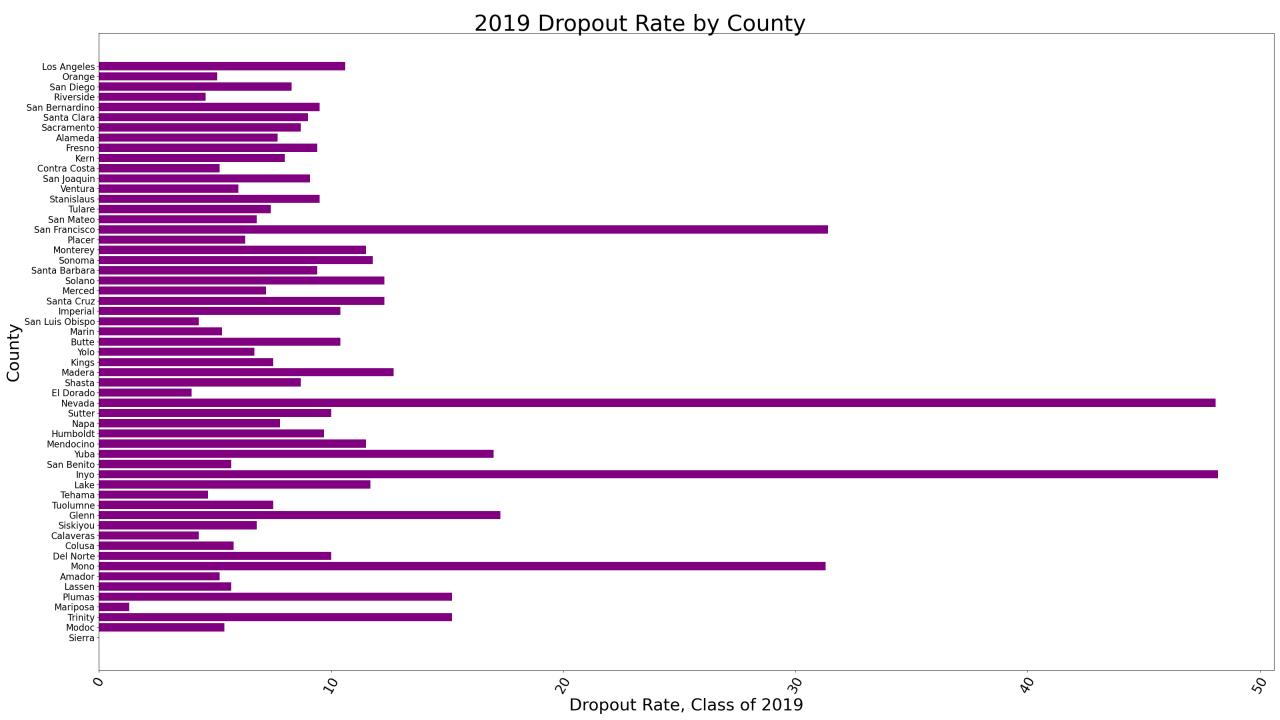
### **Graduation Cohort**

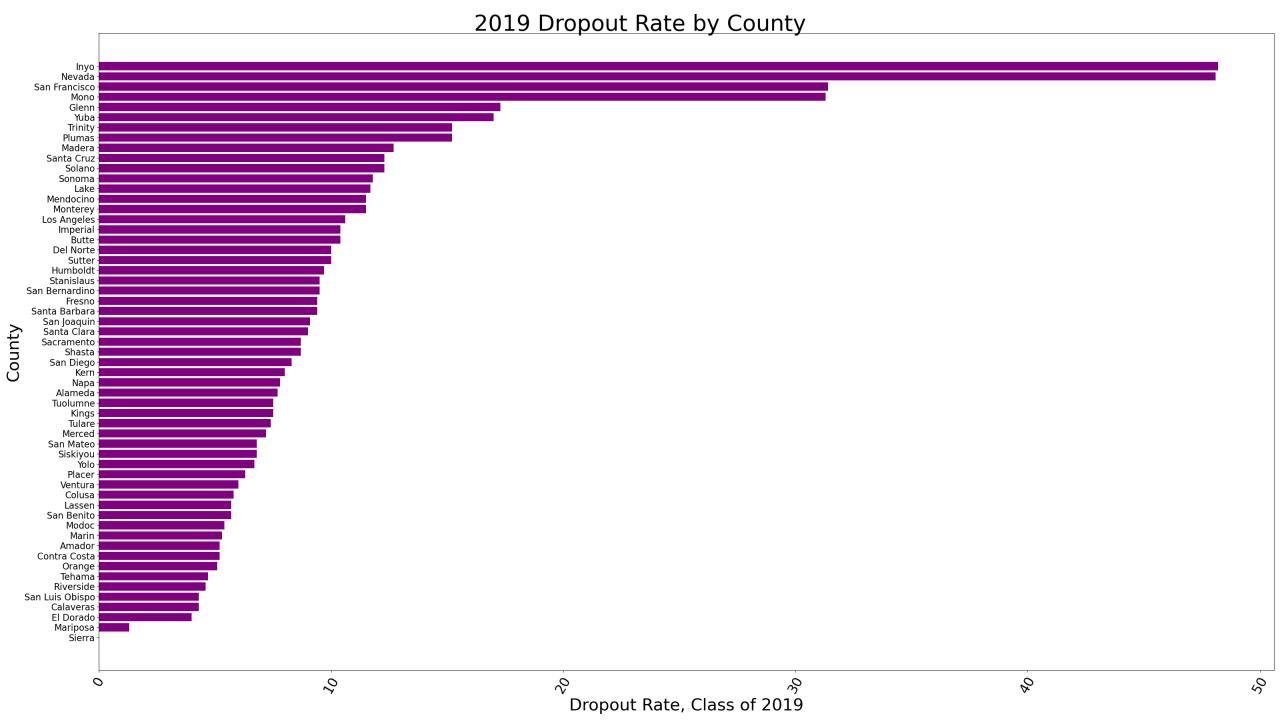
- All students who enter 9<sup>th</sup> grade at the same time
- Plus/minus students who leave or join the school
- ▶ Public school only

## Dropout Rate

- Percentage of the graduation cohort who:
- Do not complete in any way
- Are not still enrolled







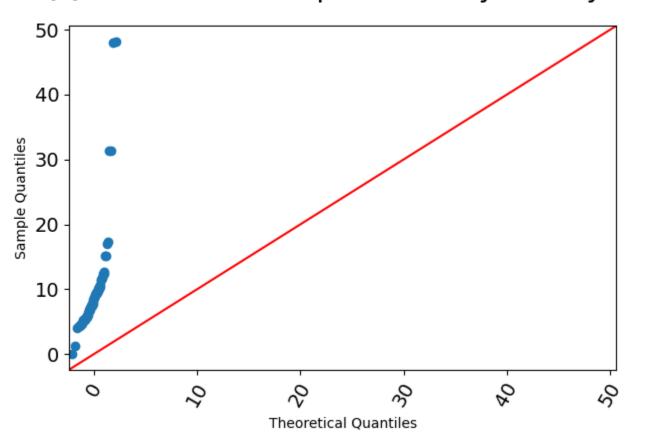
## Method

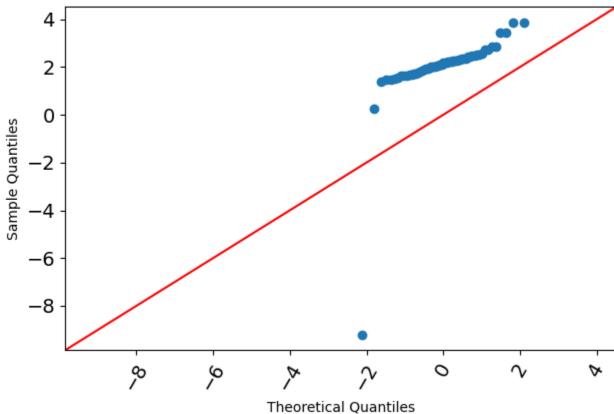
- ▶ 14 government datasets
- ▶ 1 row per county, 100s of features (at first)
- ▶ All predictors came from ~2015, dropout rate from 2019
- ▶ Feature engineering: condense, impute, polynomialize
- ▶ Feature selection: correlations, LASSO

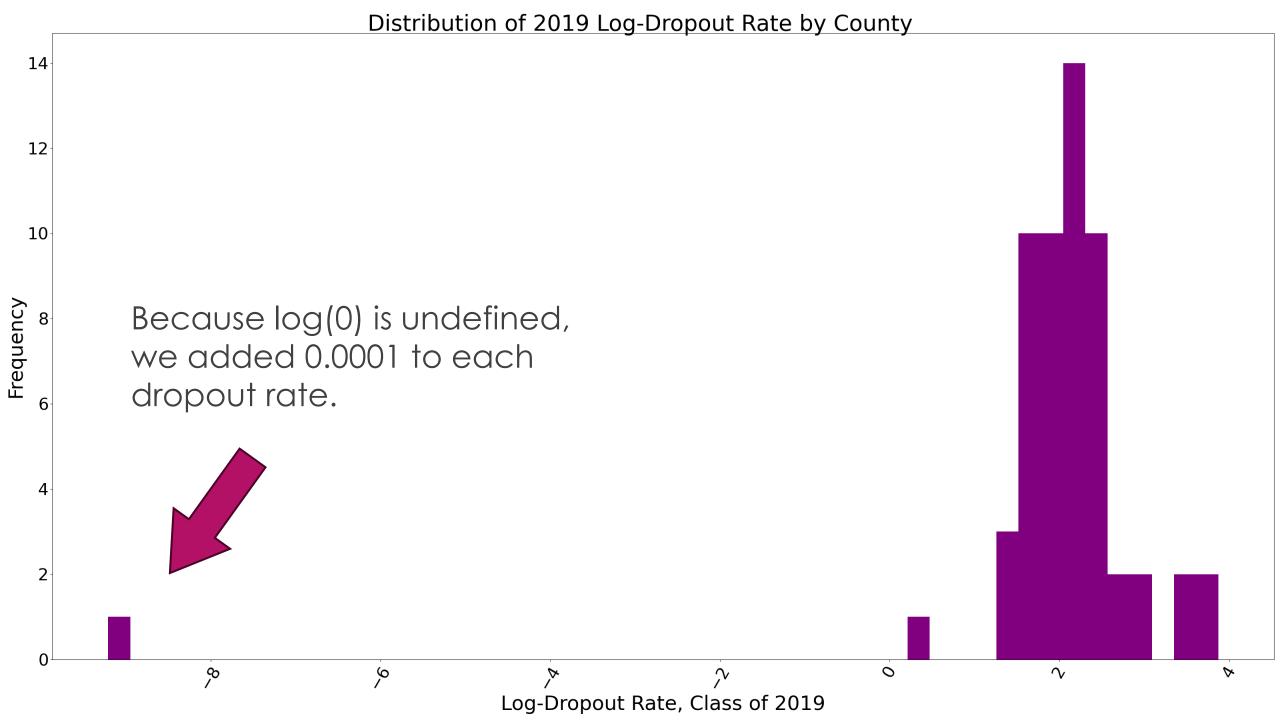
# Distribution Woes

QQ-Plot of 2019 Dropout Rate by County

QQ-Plot of 2019 Log Dropout Rate by County







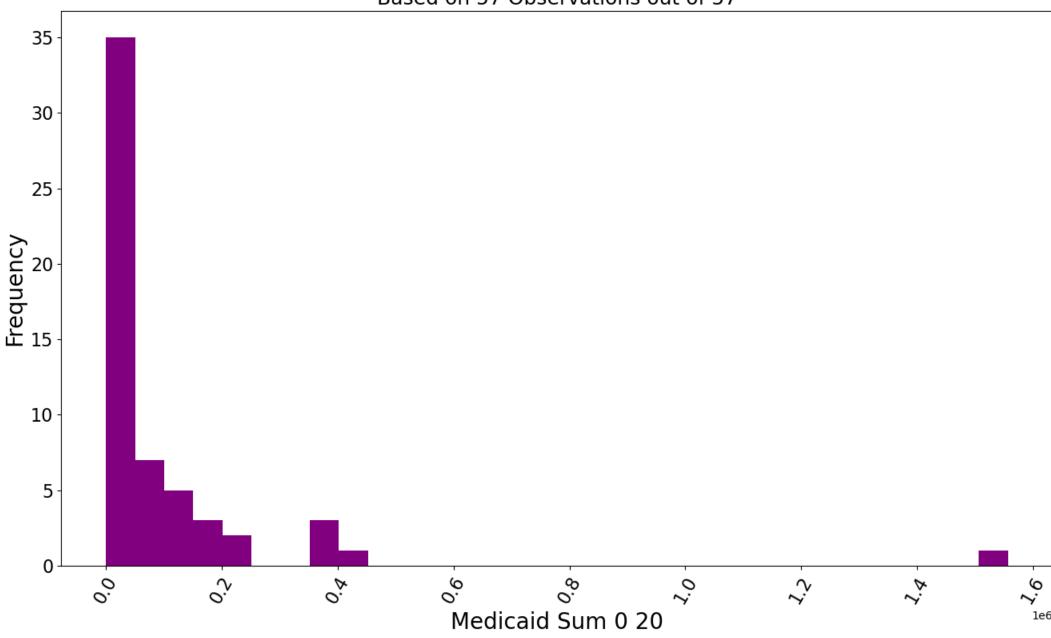
## Features

- Population & cohort size
- Poverty & income inequality
- Unemployment rates
- ► Medicaid enrollees <21yo
- ► Racial demographics <6yo
- ▶ Teenage birth rate

- Availability of ecigarettes
- ► STD rates
- Suicide, asthma death rates
- ► Infant mortality
- Medicaid abortion funding
- Daycare availability

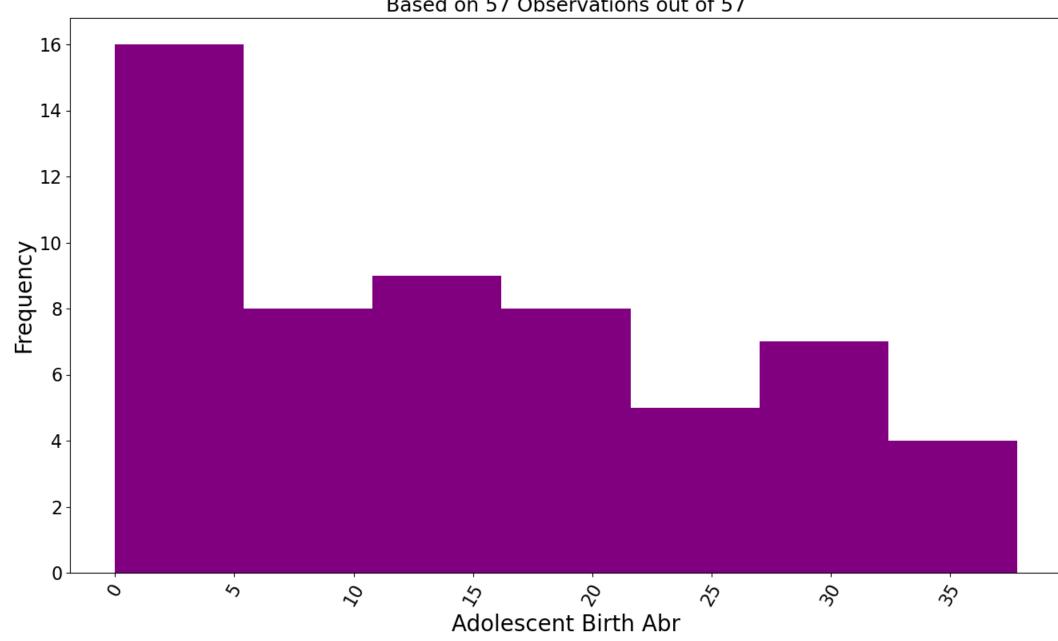
#### Distribution of Medicaid Sum 0 20





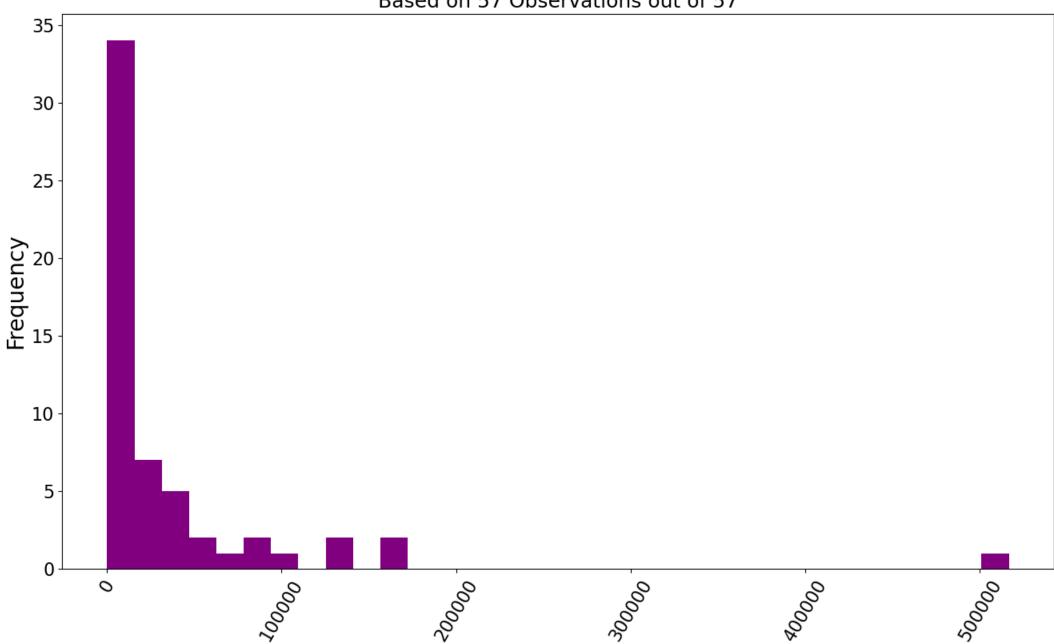
#### Distribution of Adolescent Birth Abr





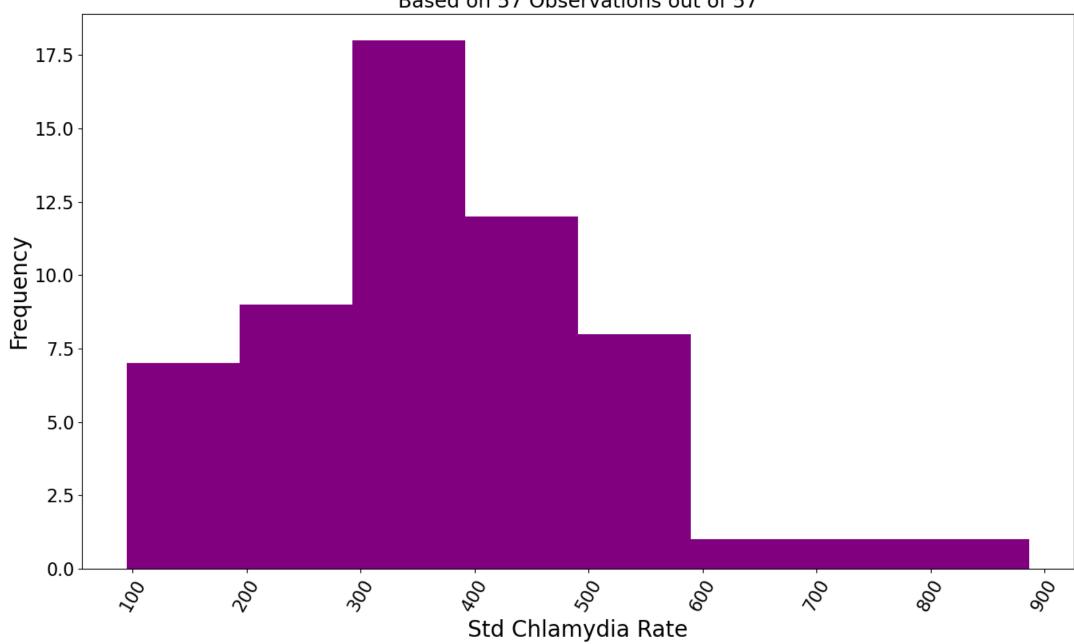
#### Distribution of Daycare Slots Child Total Pop





### Distribution of Std Chlamydia Rate





## Results

- ▶ 10 predictor variables (scaled), 3 types of models
- ▶ Regression: LASSO and Random Forest
  - ► Training: max R2 = 0.83
  - ► Testing: max R2 = 0.13
- Clustering: Kmeans
  - ▶ 13 clusters, poorly defined (silhouette = 0.24)

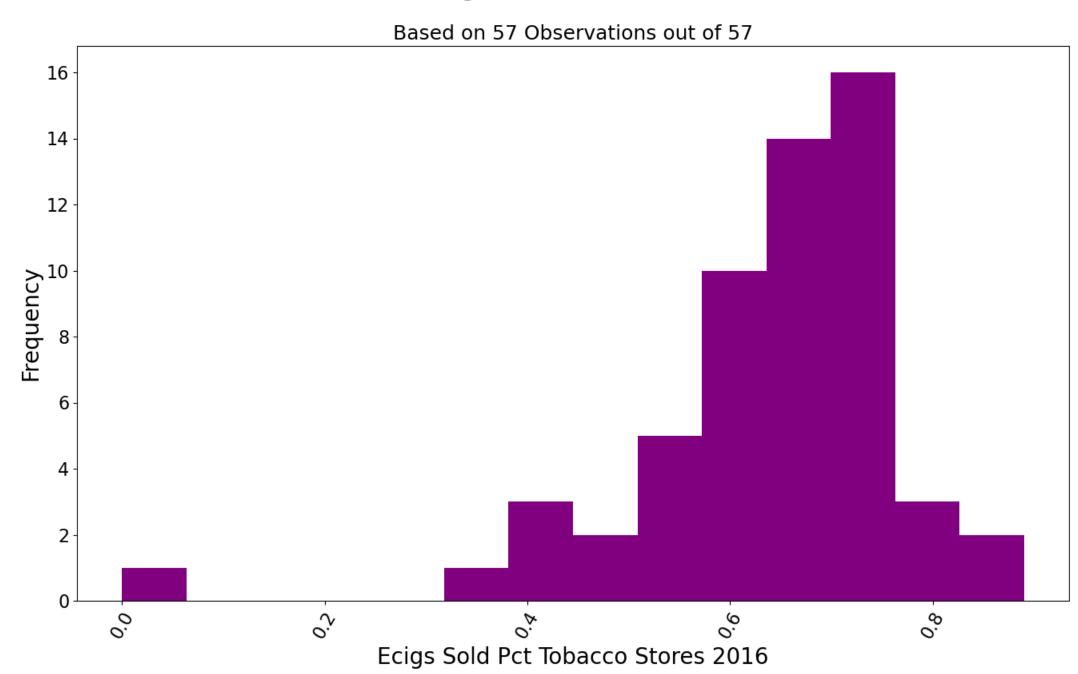
# Results

- Correlations do not imply causation, but...
- Availability of ecigarettes seems bad
- Moreso in combination with other factors than alone

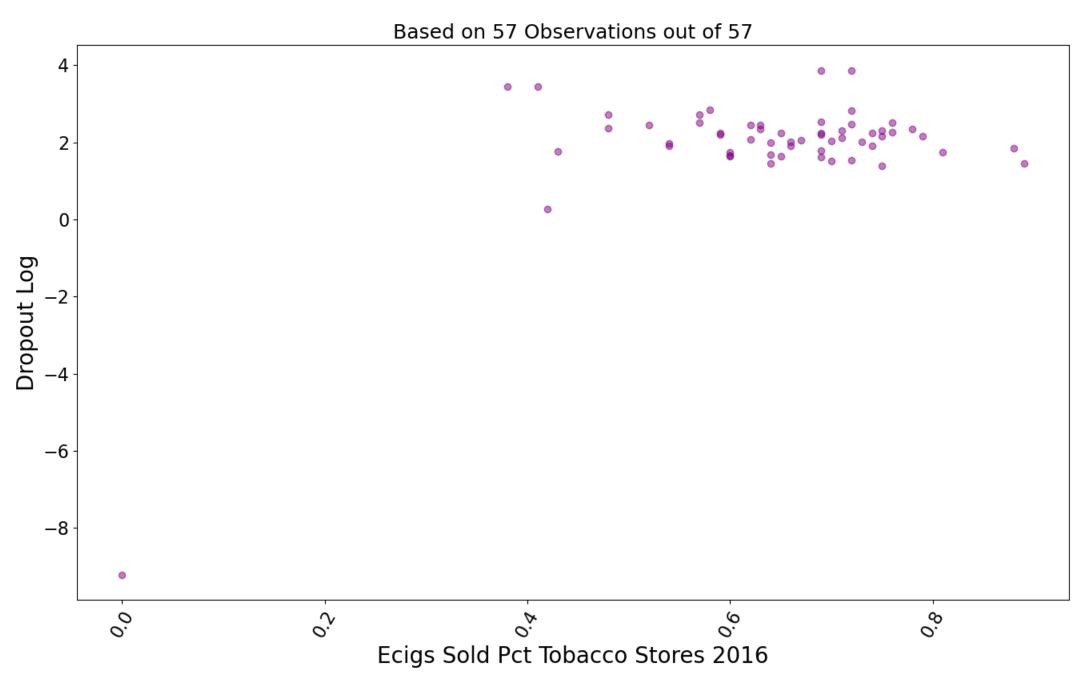


Photo credit: <u>Vaping360</u>, <u>CC BY 2.0</u>

### Distribution of Ecigs Sold Pct Tobacco Stores 2016



#### Relationship between Ecigs Sold Pct Tobacco Stores 2016 and Dropout Log



# Limitations

- Irregular data leads to faulty models
- ▶ Have to choose: exclude counties, or limit performance
- "Obvious" predictors not readily available (e.g., race)
- Many predictors available as counts, not rates
- Data is old and pre-pandemic

## Recommendations

#### For us

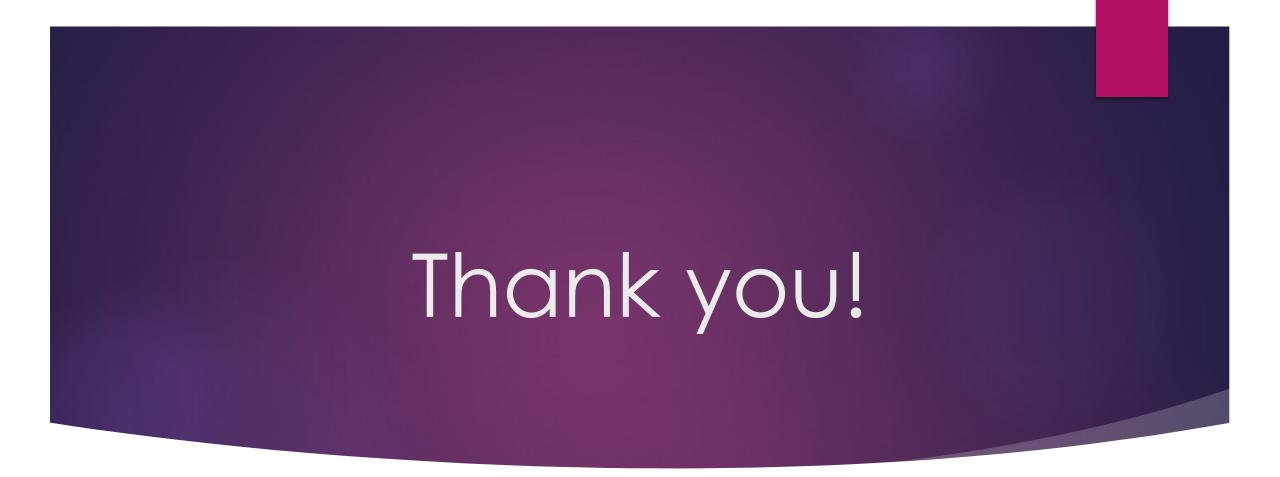
- Control for population better
- Evaluate outliers separately
- ▶ Refine, drop, add features
- ► Fresher data (Covid!)
- ▶ Change level of analysis?

## For you

- ► The problem is real, so keep using your expertise!
- Ponder combined effects
- Regulate ecigarette sales
- Evaluate capacity at city / school level

## Conclusion

- ▶ The goal was to understand and reduce dropout rates.
- Our first outing was not a huge success.
- ▶ Try to reduce the availability of ecigarettes.
  - Legal regulation
  - Cultural popularity
- ▶ It'll take more work, but we're in if you're in.



We will now take questions.