# Clustering Analysis of COVID-19 Responses by County to inform Future Pandemic Prevention

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### Context

COVID-19 has impacted people, hospitals, communities, families, and much more.

Given the available data, there is an opportunity to leverage modern machine learning to identify improvement opportunities for each county in the USA.





<sup>\*</sup>https://www.nytimes.com/interactive/2020/us/coronavirus-us-cases.html

<sup>\*\*</sup>https://ourworldindata.org/coronavirus-data

## Objectives of the study

- Identify "high-performing" groups of counties
- Execute summary statistics about why the groups might be "high-performing", and analyze "lower-performing" groups of counties to suggest improvement opportunities
- Use data-driven insights to instill new practices into how to manage COVID-19

## How is COVID control / analysis done today?

- Conservative / broad sweeping mandates to prevent transmission
- Big emphasis on cases, deaths, and vaccine doses
- Descriptive and diagnostic analytics, instead of predictive and prescriptive analytics

### Our approach

- Ingest large spectrum of data (COVID statistics, demographic, behavioral data) at county level
- > Feature engineer and normalize data
- K-Means clustering algorithm
- Visualization of K-Means clusters utilizing Principal Component Analysis (top two principal components)
- Analyze clusters using summary statistics
- Identify recommendations for each cluster to help with COVID control

### What is innovative and new?

- K-Means utilizing a broad integrated data set at county level (i.e., more granular)
- Focus on predictive / prescriptive actions for counties to take

### Who cares? What difference will this make?

#### Who Cares?

- Public officials and policymakers
- Everyday US citizens
- Pandemic researchers / scientists

#### What difference will this make?

- Potentially can save future lives
- Reduce strain on hospital / health care systems
- Mitigate negative economic impact

## Risks and payoffs



#### Risks

- Misuse of demographic data / bias
- Unreasonable recommendations



#### **Payoffs**

- Data-driven approach to improve human lives
- Reduce economic impact on society
- Reduce strain on healthcare system

### Costs and effort estimates



#### Cost

- Computing power
- Time (est. \$20,000)



#### **Effort Estimates**

- Duration: 8 weeks
- Hours: 6 resources at 20 hours per week

### Monitoring progress and success



#### Midterm check



#### Final check

- Data is available and can be integrated
- Logical groupings / clusters established
- KPIs: # of reasonable clusters, # of available features

- Reasonable summary statistics by cluster
- Reasonable / actionable recommendations by cluster to be transcended to each county
- KPIs: # of recommendations by cluster, # of next steps for study

## Effort Estimates\*/Next Steps

Name	Task(s)	Start Date	Target date
Andrew Faris	Model Dev, Report Writing, Poster Building	1/25/2022	4/23/2022
Jyothi Karra	Data Engineering, Report Writing, Poster Building	1/25/2022	4/23/2022
Morteza Maleki	Data Engineering, Report Writing, Poster Building	1/25/2022	4/23/2022
Ruchi Patel	Model Dev, Report Writing, Poster Building	1/25/2022	4/23/2022
Betsy Thorne	Visualization, Report Writing, Poster Building	1/25/2022	4/23/2022
Tony Bakshi	Model Dev, Report Writing, Poster Building	1/25/2022	4/23/2022

<sup>\*</sup>All team members have contributed a similar amount of effort.

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