Analysis of Non-Pharmaceutical Interventions (NPIs) for COVID-19 Using County-Level Mandate Data

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Background

In response to the COVID-19 pandemic, local governments had various mandated Non-Pharmaceutical Interventions (NPIs).

Dataset:

- County-level NPI data for Virginia, Minnesota, Arizona, Hawaii, New Jersey, and Washington
- Business, school, religious, and college/university closures, mask mandates, and lockdowns

Goal: Improve understanding of relationship between NPIs and COVID-related cases



Modelling

Variable Correlations:

- Number of business closures and lockdowns were fairly strong predictors of how many total mandates a county would have
- Political affiliation and income were somewhat related to total mandates implemented by a county



Model:

- Multi-output random forest classification model:
 - Predicts number of mandates implemented for a county within first 1.5 years of pandemic
 - Features: population density, percent earning below 75k/year, percent college educated, GDP, number of hospitals, number of colleges, and voting information.
 Mean absolute error: 0.49

mandate type and by state

- More restrictive mandates, like lockdowns and business closures, were shorter
- Mask mandates and school closures lasted longer overall
- Virginia and Washington had the highest average countywide mandate lengths, while Minnesota and Hawaii had the shortest.







School closures and
college closures are
somewhat correlated with
political beliefs: as
counties lean more
Republican, their mandate
length gets shorter. We
see the opposite trend for
lockdowns, though
counties that are more
educated and wealthier
tended to have longer
lockdowns.

Cases + Hospitalizations:

1. Rolling average of mandate length by start date:

This metric matches up fairly well with peaks in cases/hospitalizations, with an overall lag of about 140 days. On a statewide level, this varies from 100-180 days

Rolling Mandate Length Average (14 day) Against Hospitalizations

- Hamming loss: 0.34
- About 90% of the incorrect predictions were only off by one mandate

Variable: Variable: Variable: Variable: Variable: Variable: Variable:	FIPS pop_density pct_republican pct_under75k bachelors_pct_over25 GDP hosp_presence college_presence	Importance: Importance: Importance: Importance: Importance: Importance: Importance: Importance:	0.19 0.15 0.14 0.12 0.12 0.12 0.12 0.09 0.07
Metric: A	verage Precision	Value: 0.38	
Metric: A	verage Recall	Value: 0.43	
Metric: H	amming Loss	Value: 0.34	
Metric: M	ax depth of trees	Value: 15	
Metric: T	otal Trees	Value: 1000	
Metric: M	ean Absolute Error	Value: 0.49	

Future Work

- Further data collection:
 - Having a more robust and diverse dataset would be helpful for painting a clearer picture of the interactions between mandates, human behavior, and viral behavior
- Further predictive modelling:
 - Data is hard to collect and rapidly decaying, demonstrating the importance for accurate models
 - Further explore relationship between mandates, COVID-19, and other variables





2. Total active mandates:

Active mandates peaked at the very beginning of the pandemic and gradually declined



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