

Machine Learning Model to Inform COVID-19 Risk Mitigation Policy



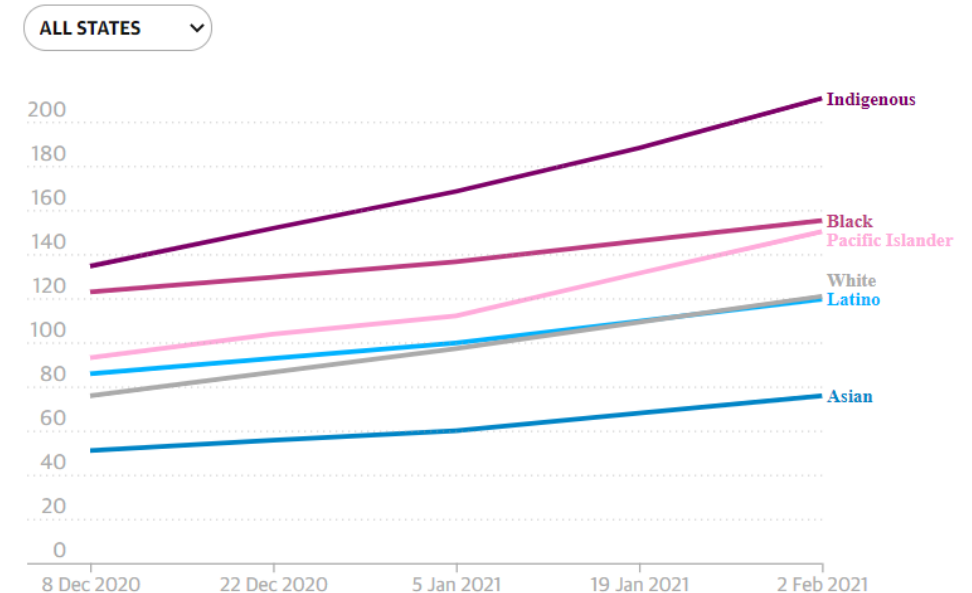
Viktor Glebov*, Gil Gallegos*, Rich Colbaugh†, Kristin Glass†
*New Mexico Highlands University, Las Vegas, New Mexico USA
†Volv Global, Lausanne, Switzerland

Background

- **Indigenous Americans** have been disproportionately hit by COVID-19, experiencing **higher rates of infection, hospitalization and death** than general population in the US
- Effective risk mitigation should be based upon analysis of the efficacy of social distancing



Covid deaths per 100,000 people, by race



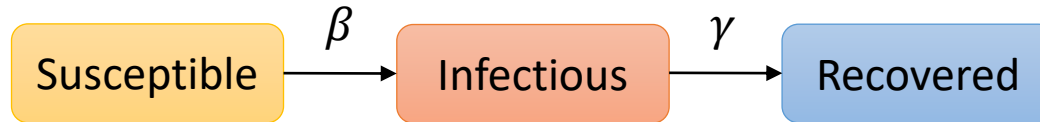
Guardian graphic. Source: APM Research Lab

Related work

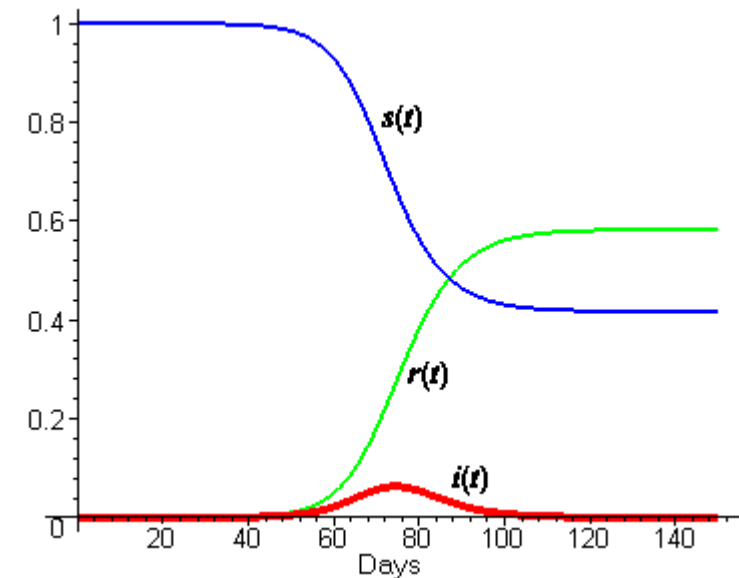
- Standard models for characterizing/predicting COVID-19 transmission:
 - include susceptible-infectious-removed (SIR) models [e.g. Chen et al. 2020, Weissman et al. 2020] and related models [e.g. Zhou et al. 2020];
 - do not yield consistently-accurate predictions or capture real-world risk mitigation policies, such as masks and social distancing.
- Machine learning (ML) models for characterizing/predicting COVID-19 transmission:
 - adopt traditional data-driven ML approach [Lalmuanawma et al. 2020, Ardabili et al. 2020, Liu et al. 2020, Pun et al. 2020, Chimmula/Zhang 2020];
 - do not leverage medical domain knowledge or nonstandard data sources.

SIR model

- Used for **mathematical modeling of spread** of infectious diseases
- Population is divided into 3 groups: **Susceptible, Infectious, Recovered**
- **Constant parameters** are used for modeling



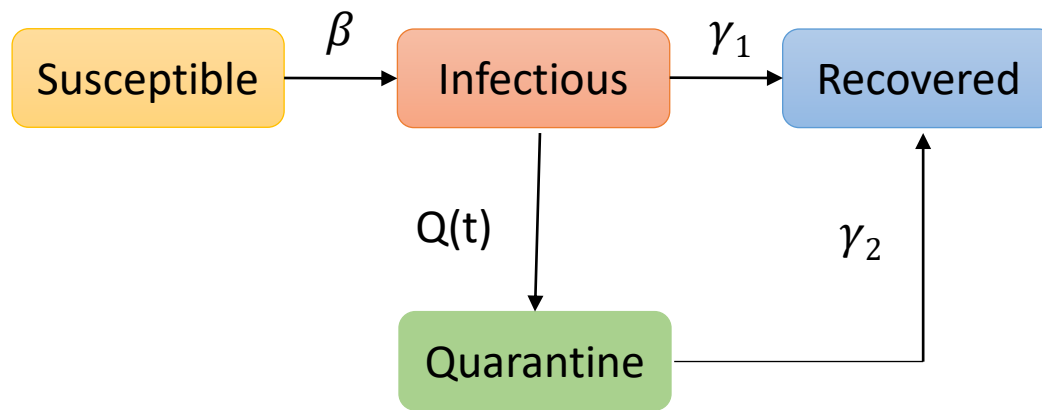
SIR model diagram



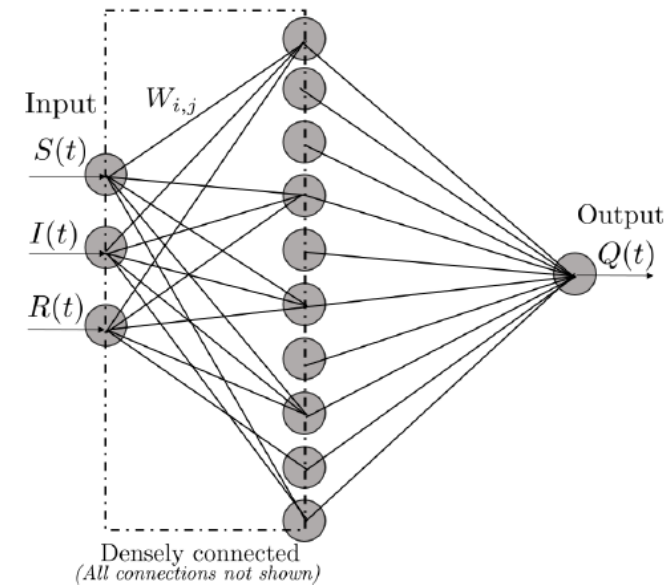
Modeling results

QSIR model

- **introduce quarantining/social distancing** into the SIR model as a function $Q(t)$
- **use the resulting Q as a metric** for public adherence to, and utility of, disease-containment policy



SIR-Q model diagram



ML model for $Q(t)$

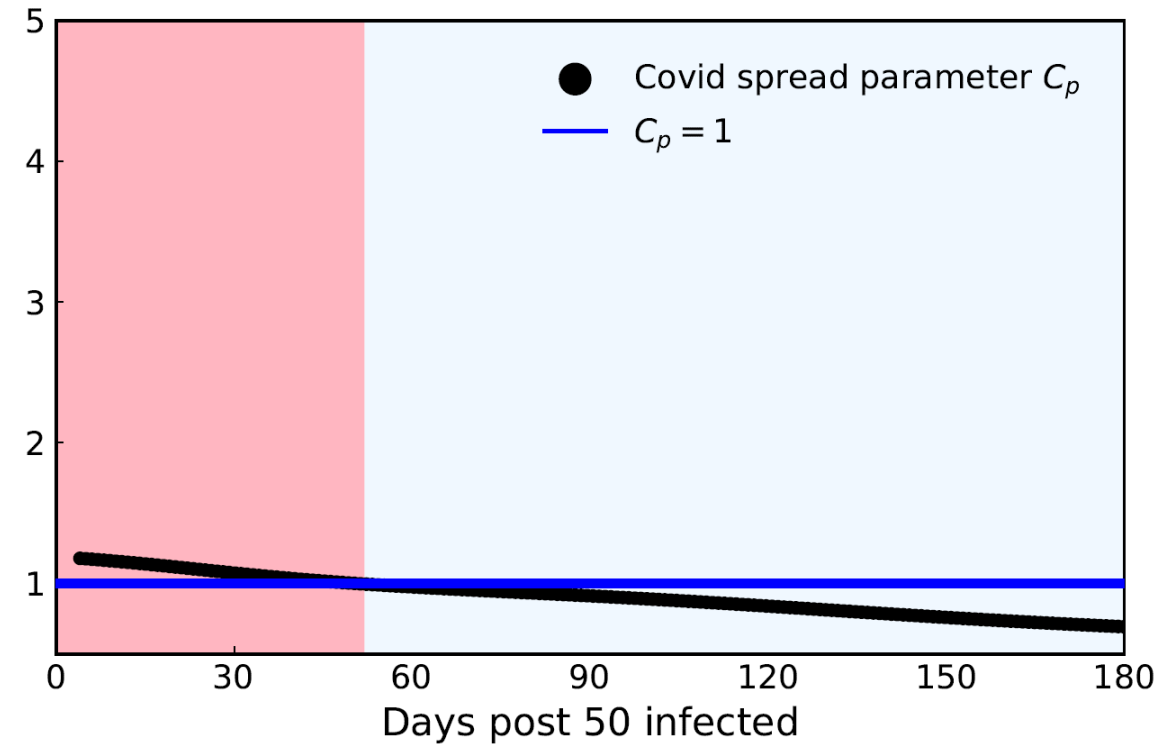
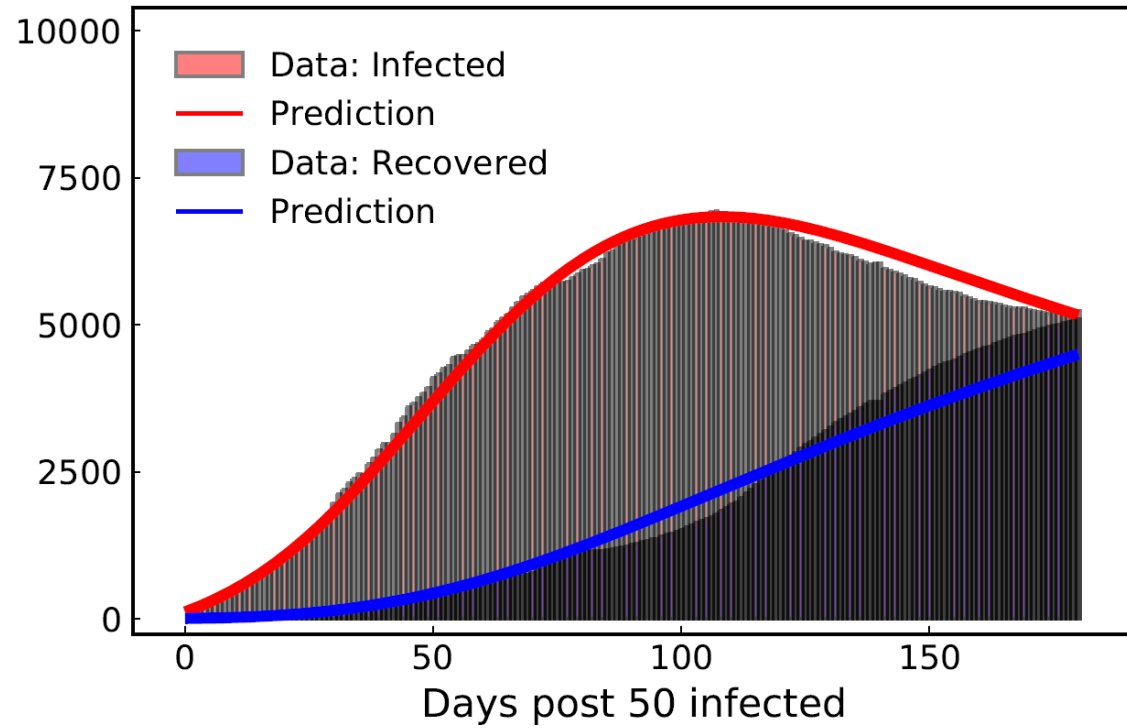
Setup

- Objective – investigate the feasibility of use of machine learning models to inform COVID-19 policy for Navajo Nation
- Tasks
 - modify and apply QSIR model to real COVID-19 data for Navajo Nation
 - estimate quarantine function $Q(t)$
 - compare results to the real data

Data sources

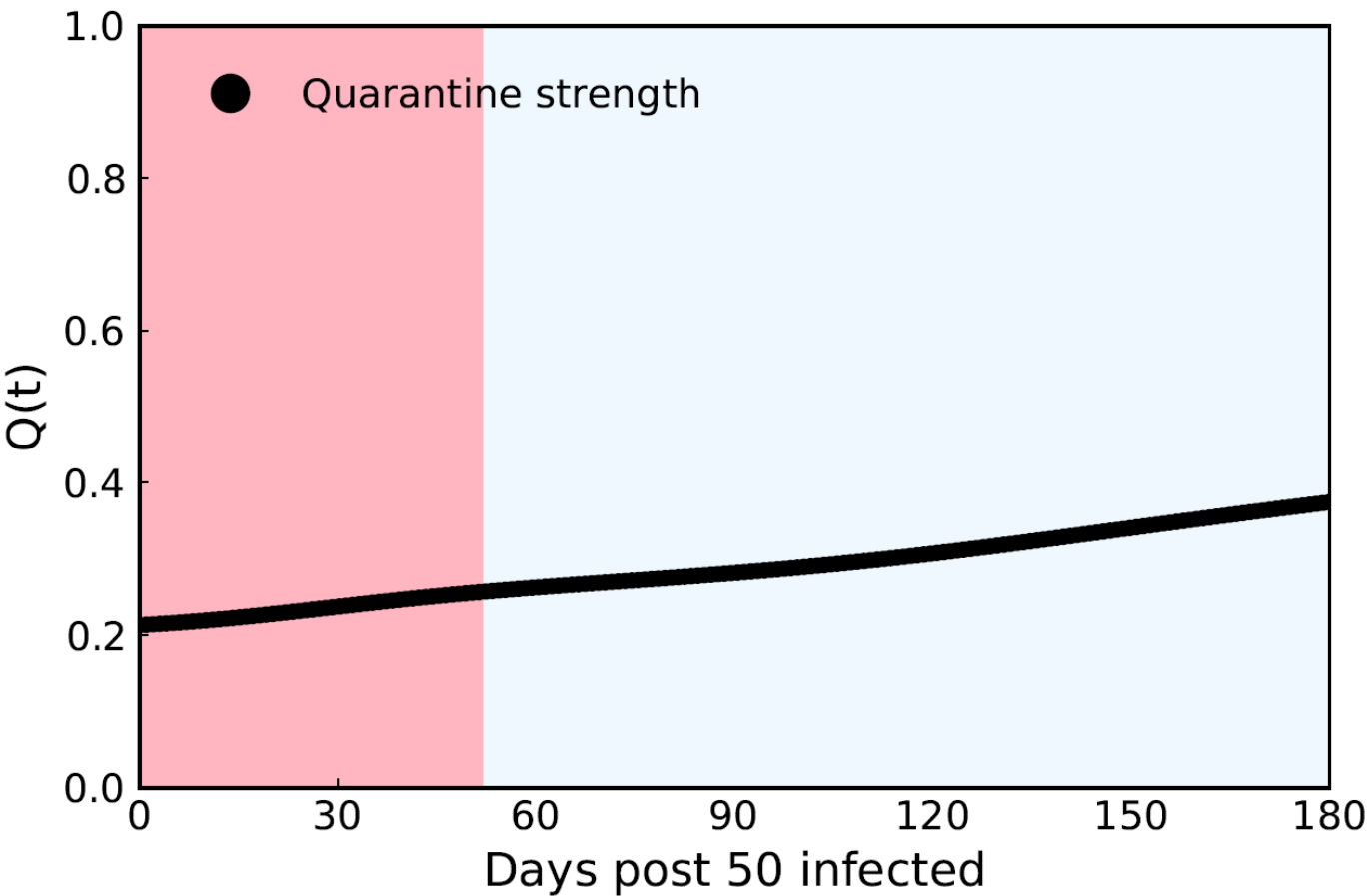
- COVID-19 infection and policy data:
 - Navajo Times <https://navajotimes.com/coronavirus-updates/covid-19-across-the-navajo-nation/>
 - Navajo Nation COVID-19 Dashboard <https://www.ndoh.navajo-nsn.gov/COVID-19/Data>
- Mobility data
 - Apple Mobility Trends <https://covid19.apple.com/mobility>
 - Google Mobility Trends <https://www.google.com/covid19/mobility/>

QSIR model

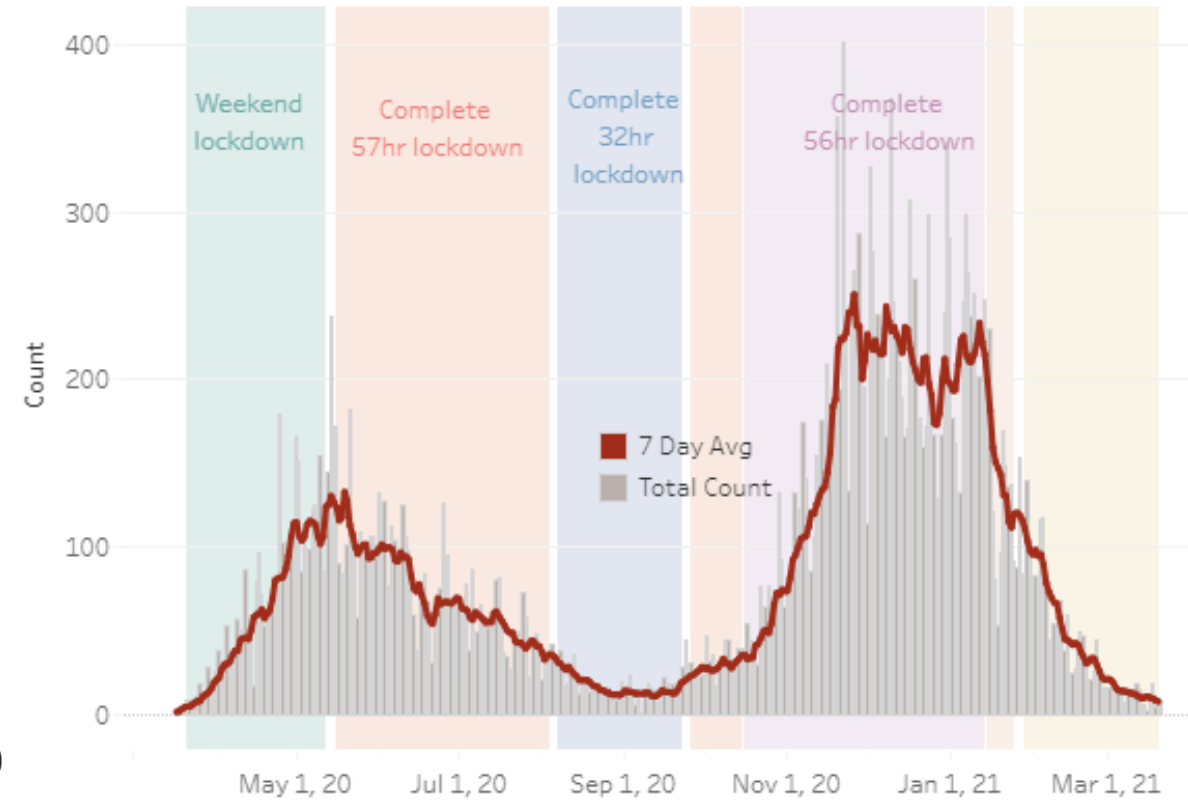


Results of modeling for the 1st wave of COVID-19 on the Navajo Nation

QSIR model validation

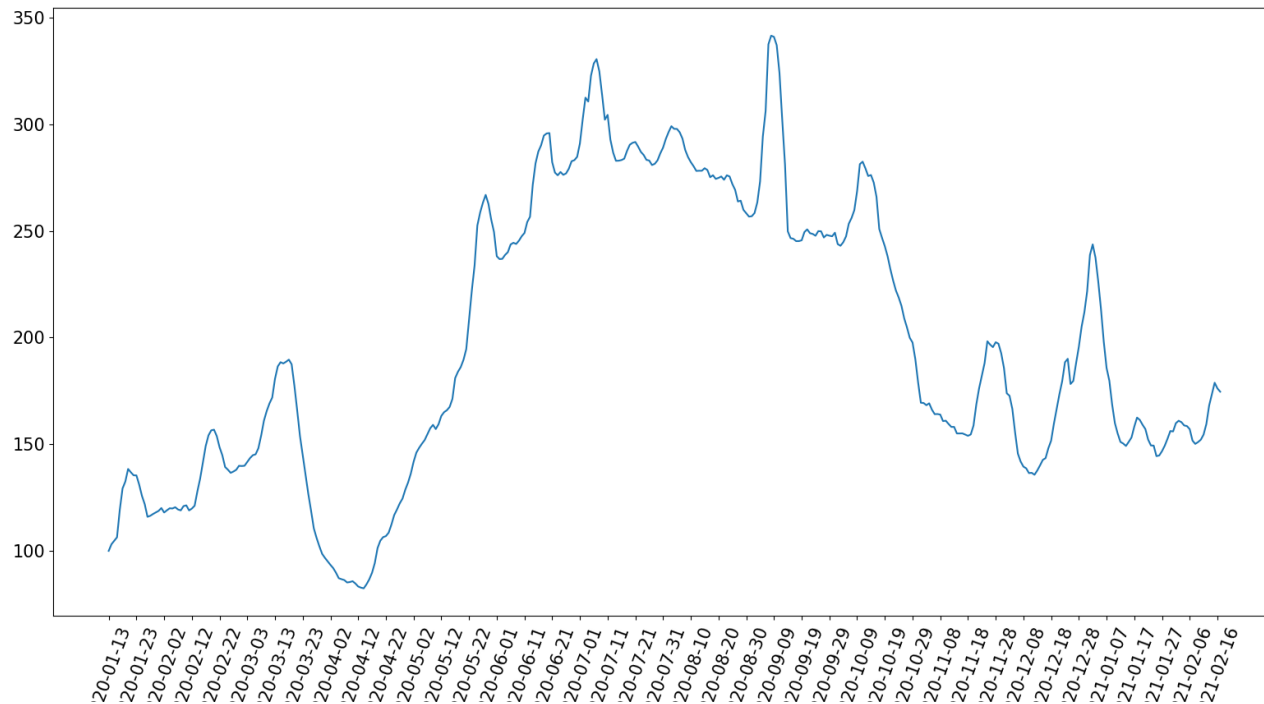
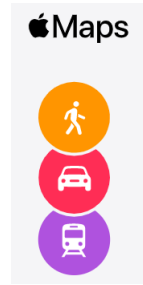


Predicted quarantine strength (1st wave)
Day 0 – April 17th



Real case count and policy
Source: Navajo Nation Department of Health

Mobility trends



Mobility trends for Navajo county

Conclusions

- Machine learning methods can improve accuracy of classical epidemiological models
- QSIR model offers potential for near real-time monitoring of generalized quarantine efficiency, a metric likely to be useful for informing/adjusting epidemic containment strategies
- There are many ways to improve/extend existing models

References

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Thank you for your attention!