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**DATAWorks 2021:
Empirical Analysis of COVID-19 in the U.S.**

Emily D. Heuring

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INSTITUTE FOR DEFENSE ANALYSES
4850 Mark Center Drive
Alexandria, Virginia 22311-1882



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About This Publication

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For more information:

Emily Heuring, Project Leader
eheuring@ida.org • 703-845-4374

Robert R. Soule, Director, Operational Evaluation Division
rsoule@ida.org • (703) 845-2482

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**DATAWorks 2021:
Empirical Analysis of COVID-19 in the U.S.**

Emily D. Heuring

Executive Summary

The zoonotic emergence of the coronavirus SARS-CoV-2 at the beginning of 2020 and the subsequent global pandemic of COVID-19 has caused massive disruptions to economies and health care systems, particularly in the United States. This briefing to be presented at DATAWorks 2021 will describe IDA's empirical analysis of COVID-19 data within the U.S. general population.

Using the results of serology testing, we have developed true prevalence estimates for COVID-19 case counts in the U.S. over time, which allows for more precise estimates of infection and case fatality rates throughout the course of the pandemic. To elucidate the behavioral, demographic, weather, and policy factors that contribute to or inhibit the spread of COVID-19, IDA compiled panel data sets of empirically derived, publicly available COVID-19 data. We then analyzed which factors were most highly correlated with both the increased and decreased spread within U.S. states and counties. Our analysis shows that mobility to retail and recreation locations (i.e., bars and restaurants) was highly correlated with increased COVID-19 case counts.

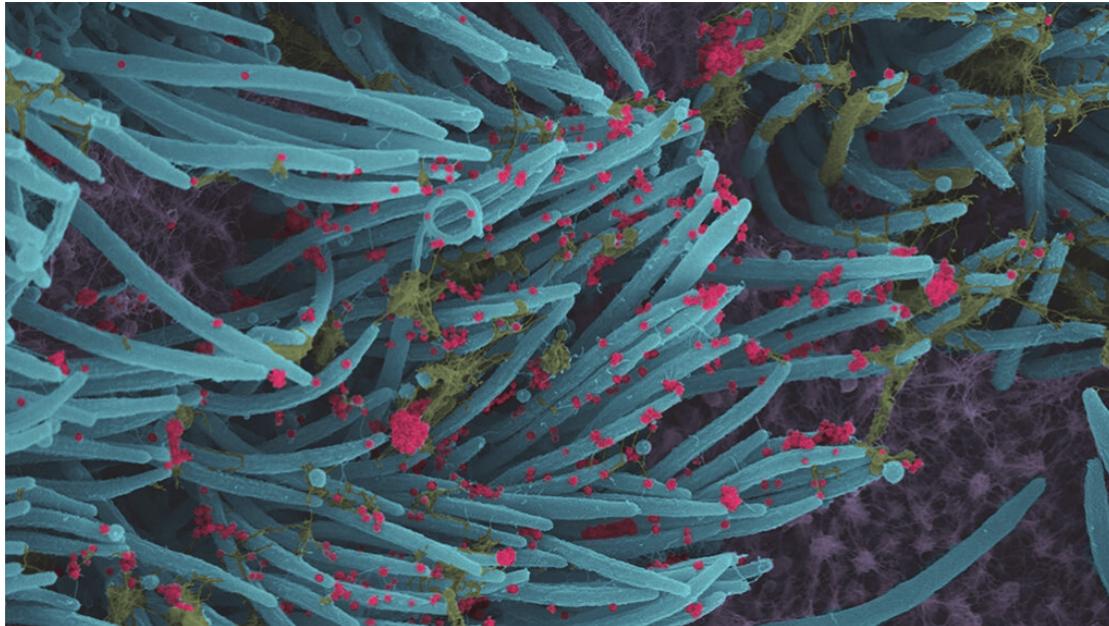
SARS-CoV-2 genomic analysis over the course of the pandemic demonstrates that different lineages have emerged and waned over the course of the pandemic.

Overall, the emergence of SARS-CoV-2 and COVID-19 has resulted in the public having unparalleled access to data for analyzing the pandemic. Such access has allowed for widespread real-time analysis to support the deployment of resources and implementation of policy. However, along with this unparalleled access to data come potential problems, as researchers may lack insight into data collection methods and experience with new types of data, which can lead to conflicting interpretations and conclusions about the pandemic.



Empirical Analysis of COVID-19 in the U.S.

Dr. Emily Heuring



Institute for Defense Analyses

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Early in the pandemic, the sponsor identified a need for empirical data and analysis to support decision making

- Traditional epidemiology models (SEIR) show what can happen, not what will happen
 - Changes in behavior change the trajectory of the spread of the virus
- Sponsor identified a need to compile a large data set of empirical data from a variety of sources
- It is important to identify the factors most associated with both mitigation and spread of the virus

COVID-19 is the first pandemic in history in which vast amounts of data are openly accessible to the public

IPUMS USA
 HOME | SELECT DATA | MY DATA
 U.S. CENSUS DATA FOR SOCIAL, ECONOMIC, AND HEALTH RESEARCH
 USE IT FOR GOOD -- NEVER FOR EVIL

World Health Organization
COVID Data Tracker

SAFE GRAPH | U.S. Consumer Activity During COVID-19 Pandemic

BLAVATNIK SCHOOL OF GOVERNMENT | **UNIVERSITY OF OXFORD**
 Research > Research projects > COVID-19 Government Response Tracker...
COVID-19 GOVERNMENT RESPONSE TRACKER
 Governments are taking a wide range of measures in response to the COVID-19 outbreak. This tool aims to track and compare policy responses around the world, rigorously and consistently.

Google COVID-19 Community Mobility Reports
 See how your community is moving around differently due to COVID-19
 As global communities respond to COVID-19, we've heard from public health officials that the same type of aggregated, anonymized insights we use in products such as Google Maps could be helpful as they make critical decisions to combat COVID-19.
 These Community Mobility Reports aim to provide insights into what has changed in response to policies aimed at combating COVID-19. The reports chart movement trends over time by geography, across different categories of places such as retail and recreation, groceries and pharmacies, parks, transit stations, workplaces, and residential.

COVID-19 Dashboard by JHU
 Global Cases: 117,250,914
 2,604,123 364,845,417

NOAA NATIONAL CENTERS FOR ENVIRONMENTAL INFORMATION
 Formerly the National Climatic Data Center (NCDC)...
 Global Historical Climate Network Daily - Description

Nextstrain
 Real-time tracking of pathogen evolution
 Nextstrain is an open-source project to harness the scientific and public health potential of pathogen genome data. We provide a continually-updated view of publicly available data alongside powerful analytic and visualization tools for use by the community. Our goal is to aid epidemiological understanding and improve outbreak response. If you have any questions, or simply want to say hi, please give us a shout at hello@nextstrain.org.

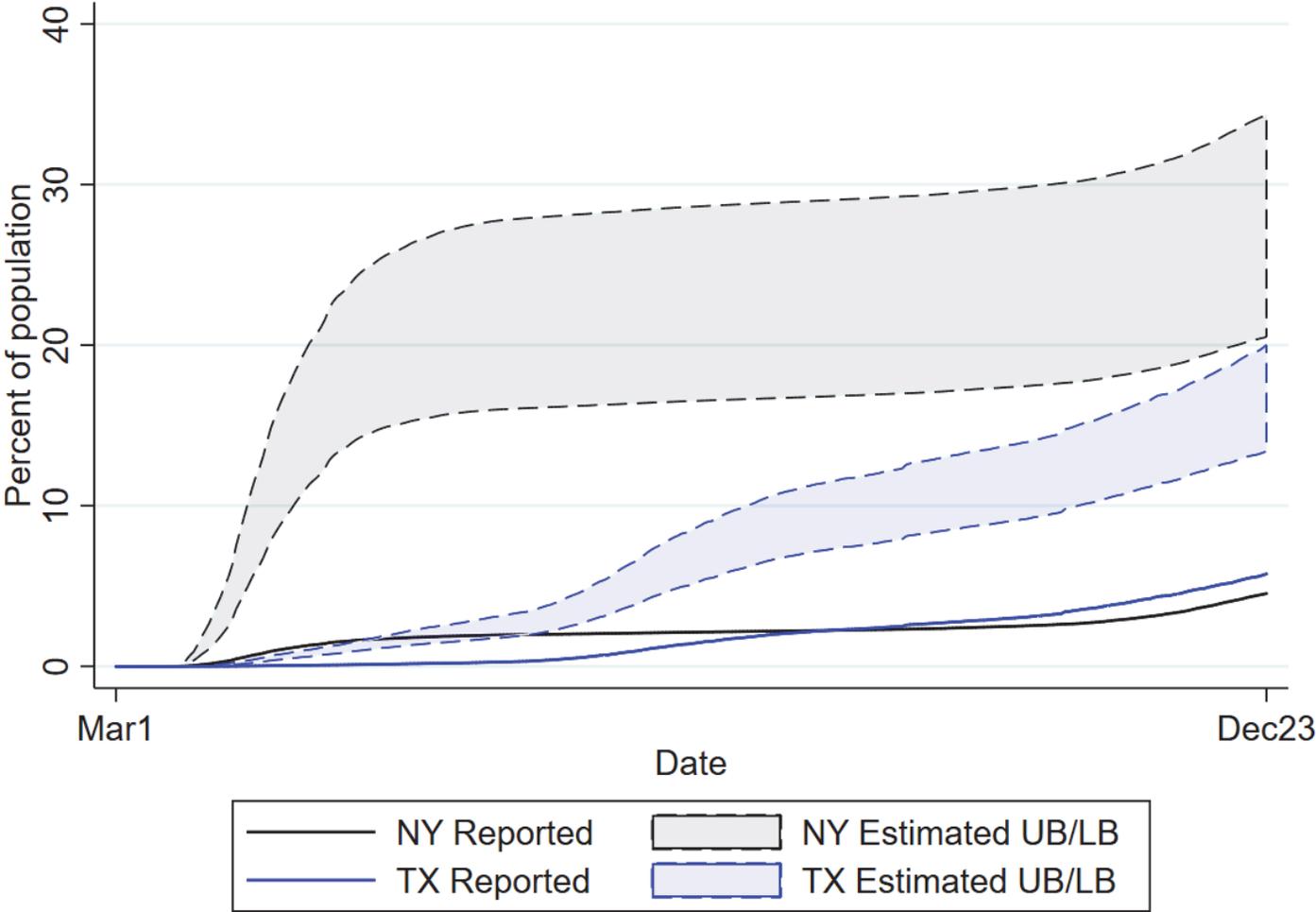
True Prevalence of COVID-19 vs Reported Case Counts

COVID-19 case counts have been chronically underreported from the beginning of the pandemic

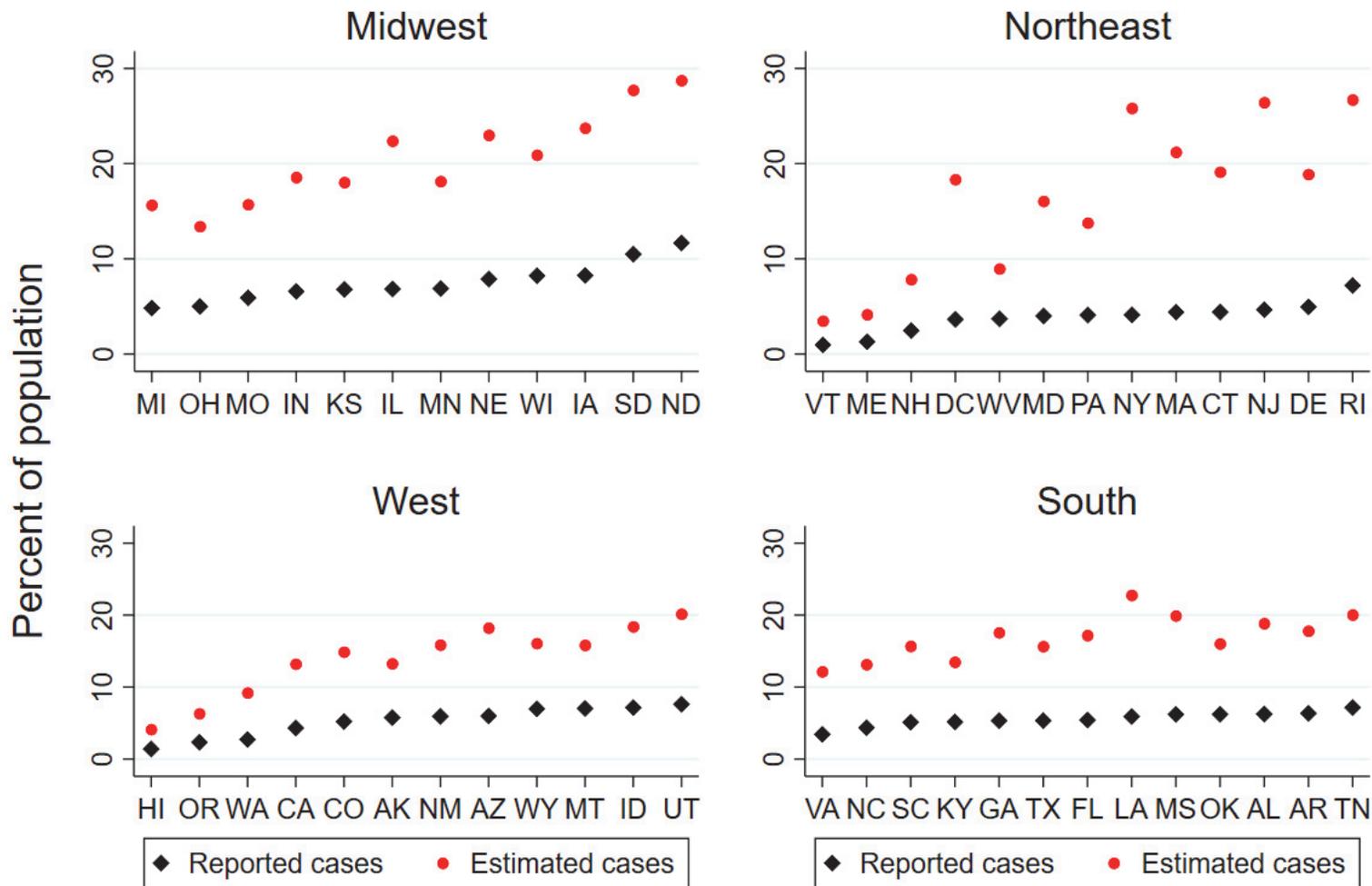
- Causes of underreporting:
 - Inadequate testing supplies and access during initial months
 - Those who had mild/asymptomatic cases weren't aware that they had COVID-19 and did not get tested
- We use serology data to estimate true COVID-19 prevalence in the U.S.
 - Serology testing detects antibodies to SARS-CoV-2 from previous infection
 - Although many city and state-level seroprevalence estimates were published, many had biased sampling techniques
 - We prefer CDC's seroprevalence methods: <https://www.cdc.gov/coronavirus/2019-ncov/cases-updates/geographic-seroprevalence-surveys.html>
- Compare the antibody prevalence to the number of reported cases at multiple points in time
 - Ten sample locations initially (has since been expanded)
 - Assume that for a given day, all regions in the U.S. had the same underreporting ratio
 - Assume no antibody decay (conflicting data in the literature)

True prevalence estimates create a clearer picture of the pandemic

Estimated vs Reported Share of Population Infected
Cumulative



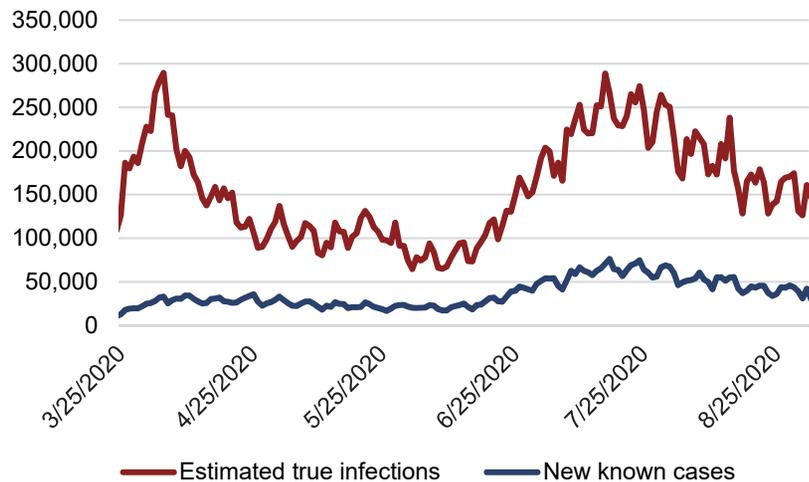
Using seroprevalence data, we can estimate the percentage of the population in each state who have been infected with COVID-19



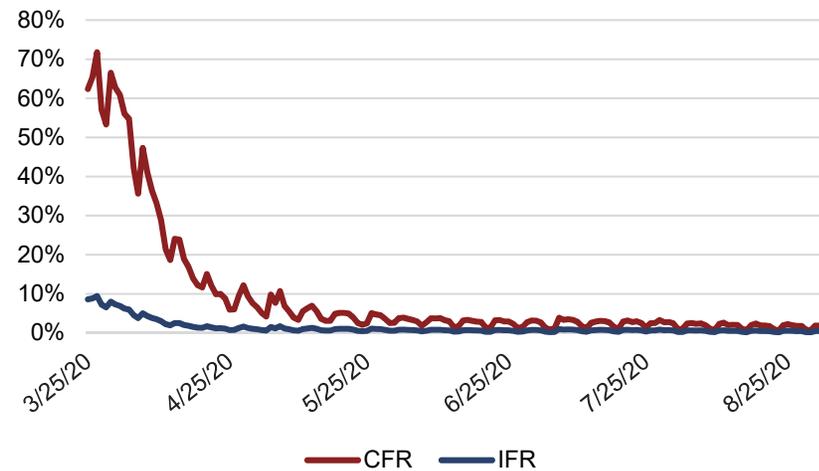
Data as of December 16

With true prevalence estimates, we can calculate infection fatality rates for the U.S.

Estimated True Infections and Known Cases



Case Fatality Rate (CFR) and Infection Fatality Rate (IFR)



Period-average fatality rate	CFR	IFR
Mar 26-Apr 30	31.56%	3.69%
May 1-31	5.07%	0.92%
Jun 1-30	2.79%	0.62%
Jul 1-31	2.53%	0.65%
Aug 1-Sep 8	1.66%	0.46%
Total deaths/total cases or infections during Mar 6-Sep 2	2.89%	0.69%

Effect of mitigation policies on the spread of COVID-19

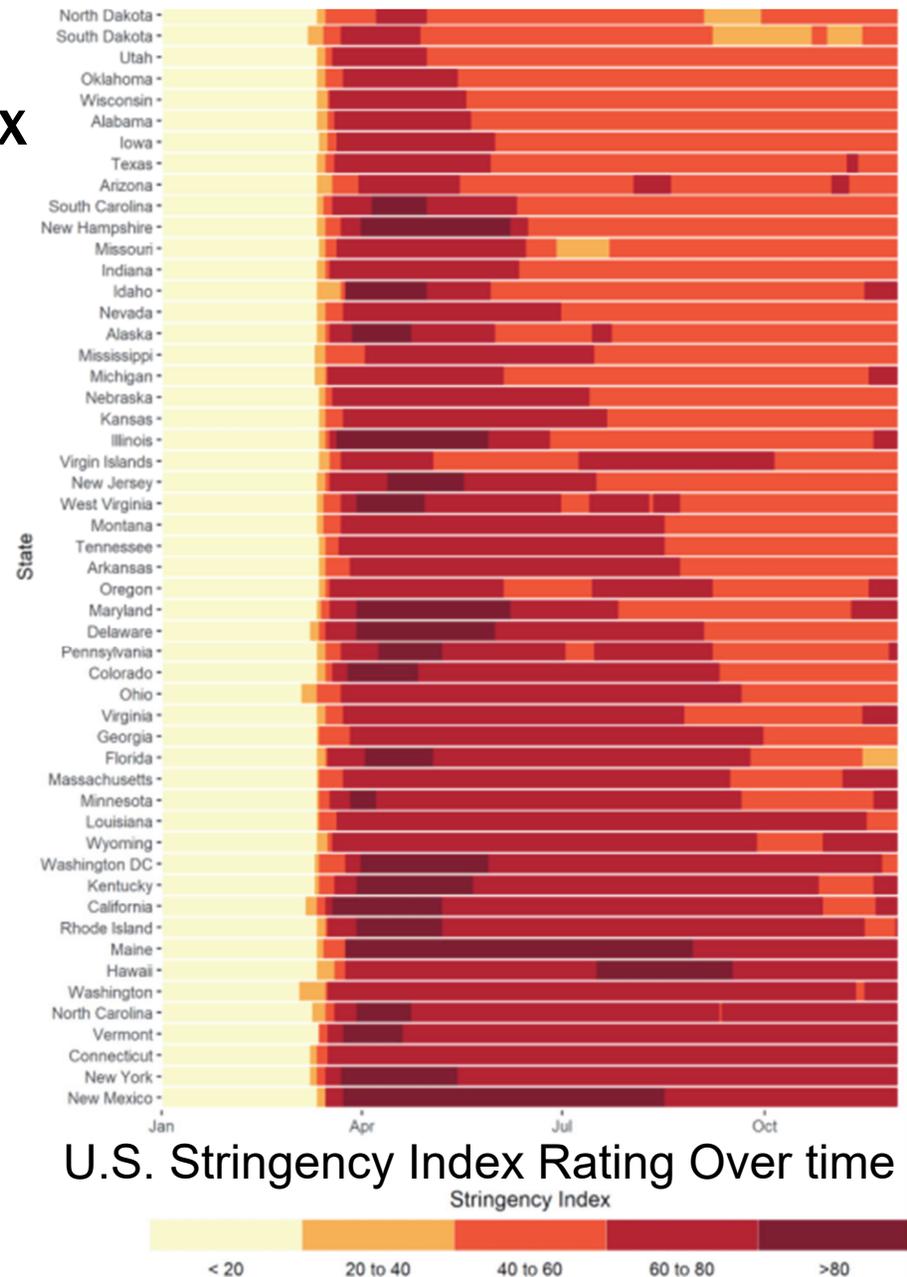
We examined how mitigation policies over the course of the pandemic affected COVID-19 case growth or decline

- To reduce bias from many mitigation policies being implemented simultaneously, we use a two-stage approach:
 - How policies affect mobility (first stage)
 - How changes in mobility from policies affect new cases (second stage)
- We find evidence that policies aimed at retail and recreation traffic decrease new cases

Example policy data – Oxford’s Policy Stringency Index

Oxford US State Policy Stringency Index

Measure	Name	Type	Targeted/general?
Containment and closure			
C1	School closing	Ordinal	Geographic
C2	Workplace closing	Ordinal	Geographic
C3	Cancel public events	Ordinal	Geographic
C4	Restrictions on gathering size	Ordinal	Geographic
C5	Close public transport	Ordinal	Geographic
C6	Stay-at-home requirements	Ordinal	Geographic
C7	Restrictions on internal movement	Ordinal	Geographic
C8	Restrictions on international travel	Ordinal	No
Health systems			
H1	Public information campaign	Ordinal	Geographic

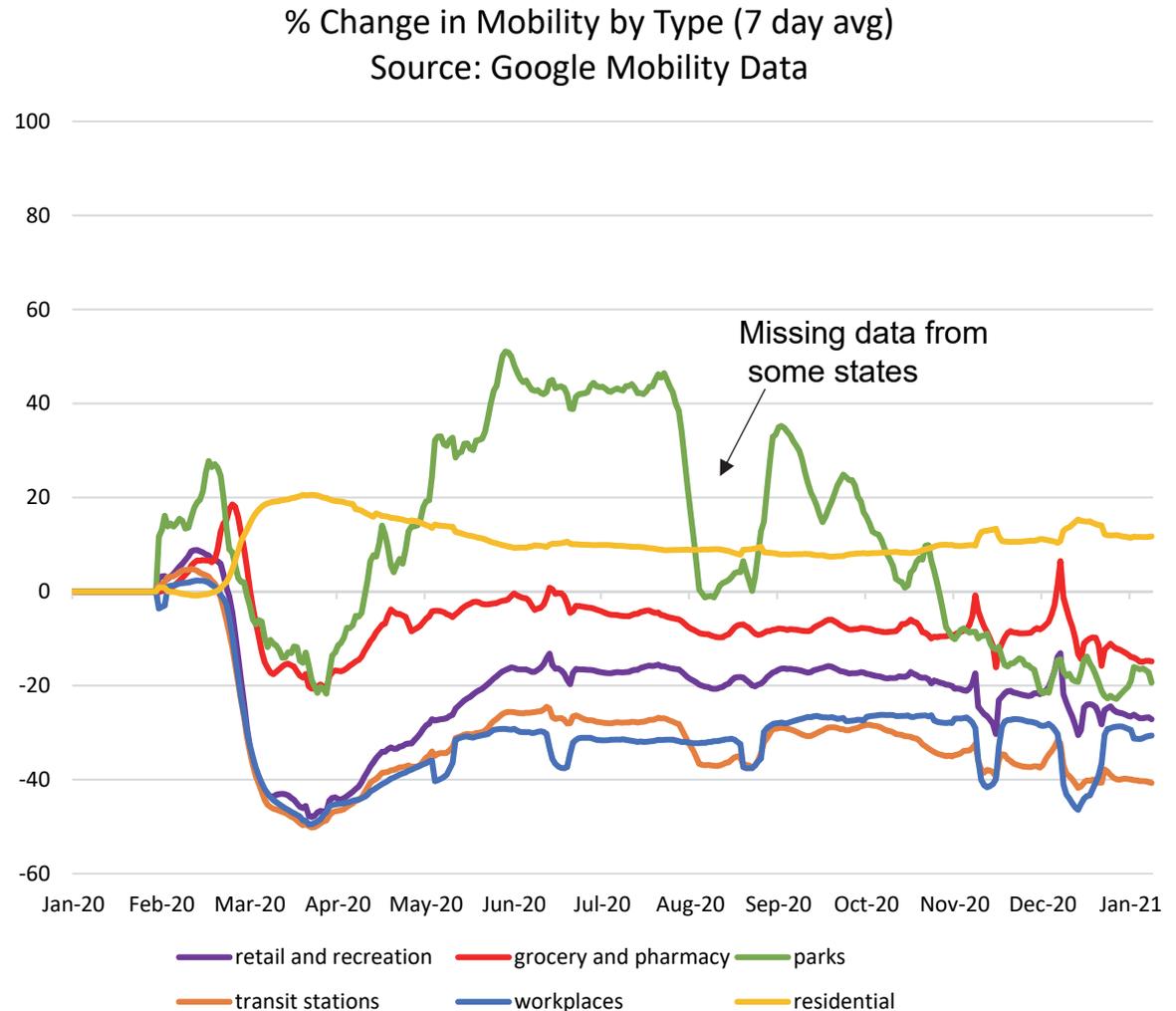


U.S. Stringency Index Rating Over time



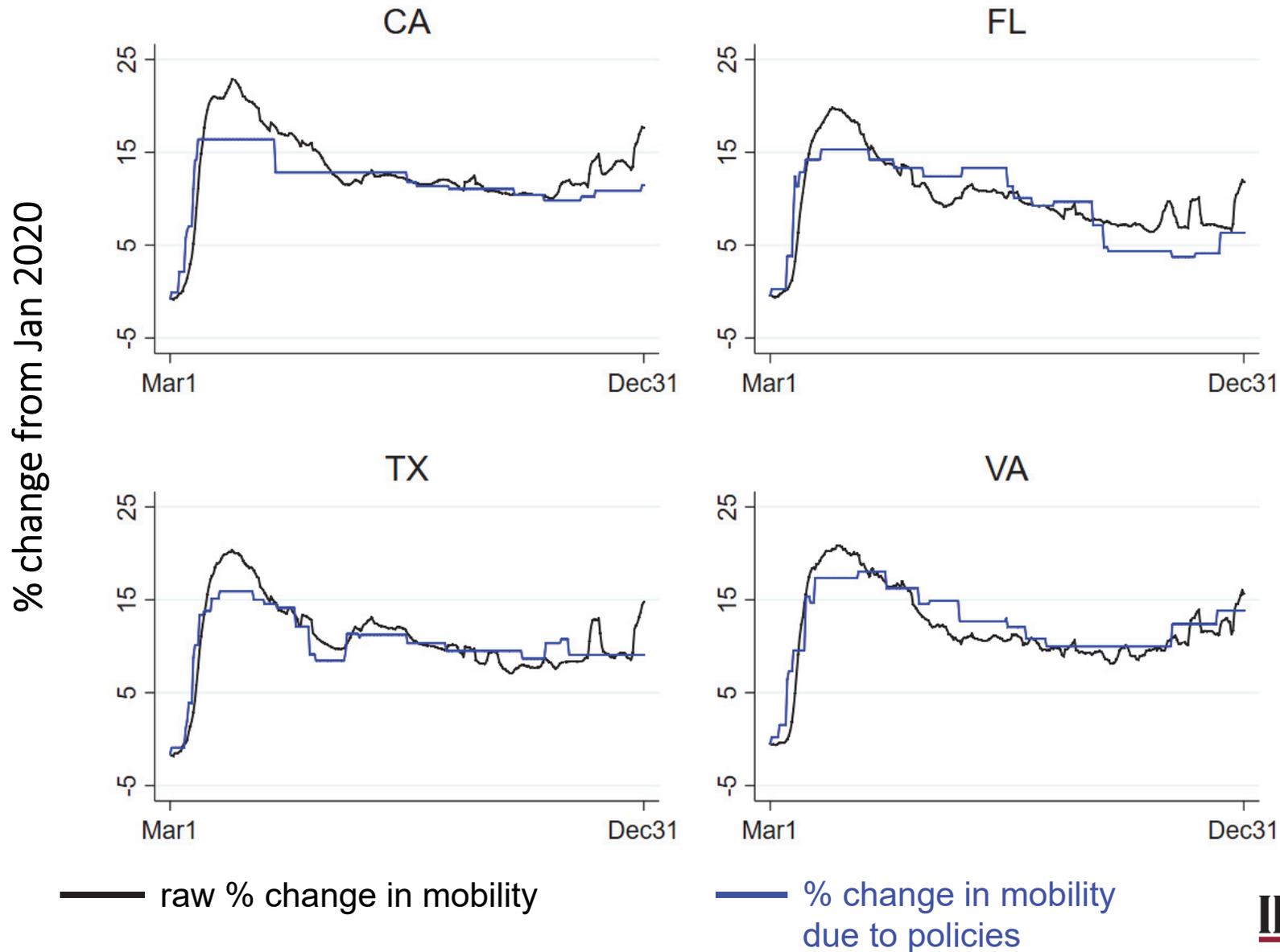
In the U.S. in 2020, mobility to most places fell below baseline except in parks and residential locations

- Mobility to most places was below the baseline for most of 2020
- Parks and residential mobility was above baseline as people stayed home or went outside
- Gradual decline at end of CY20/early CY21 as cases increased

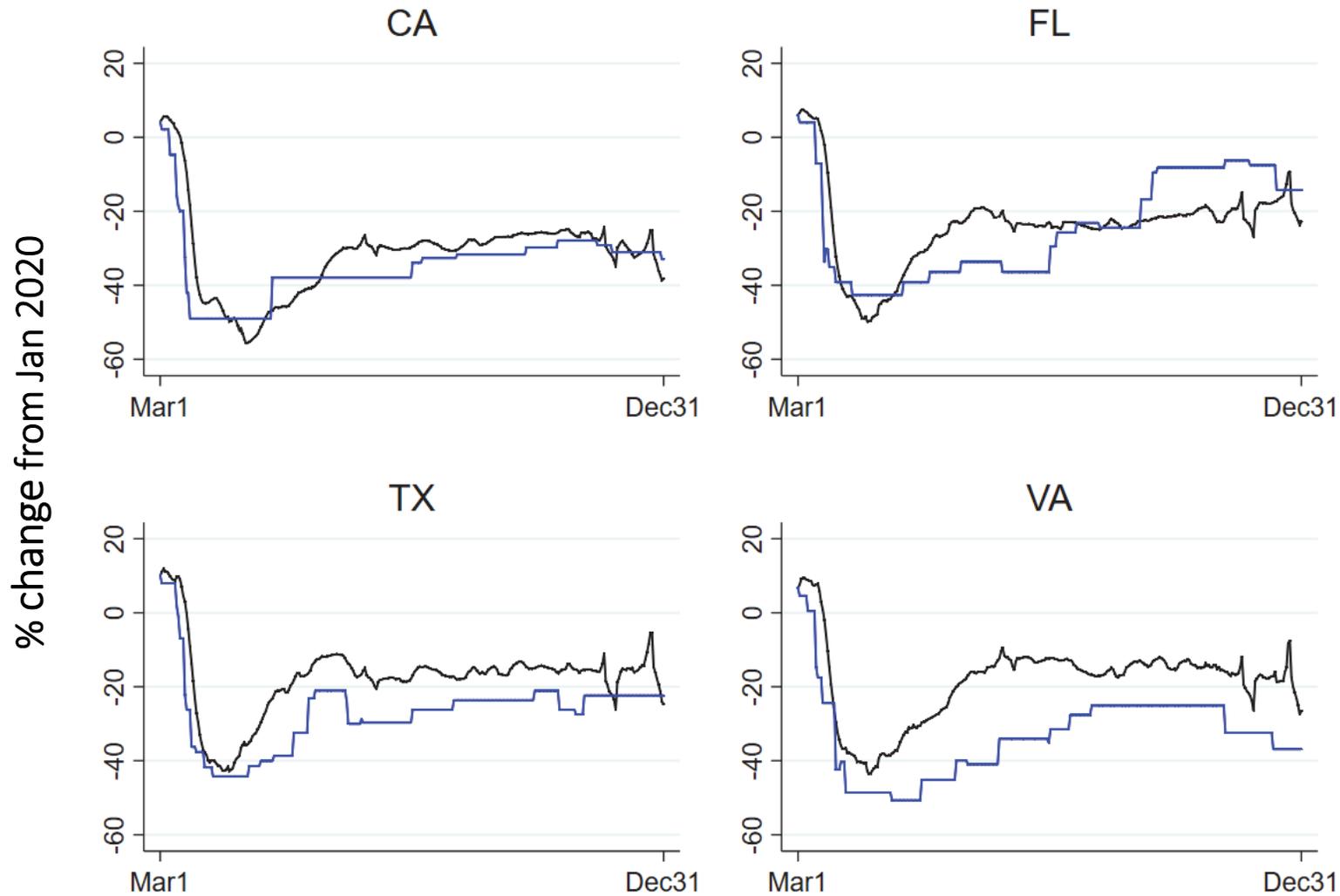


Use cell-phone mobility data as a proxy for adherence to mitigation policies

Mitigation policy stringency is strongly correlated with increased residential mobility



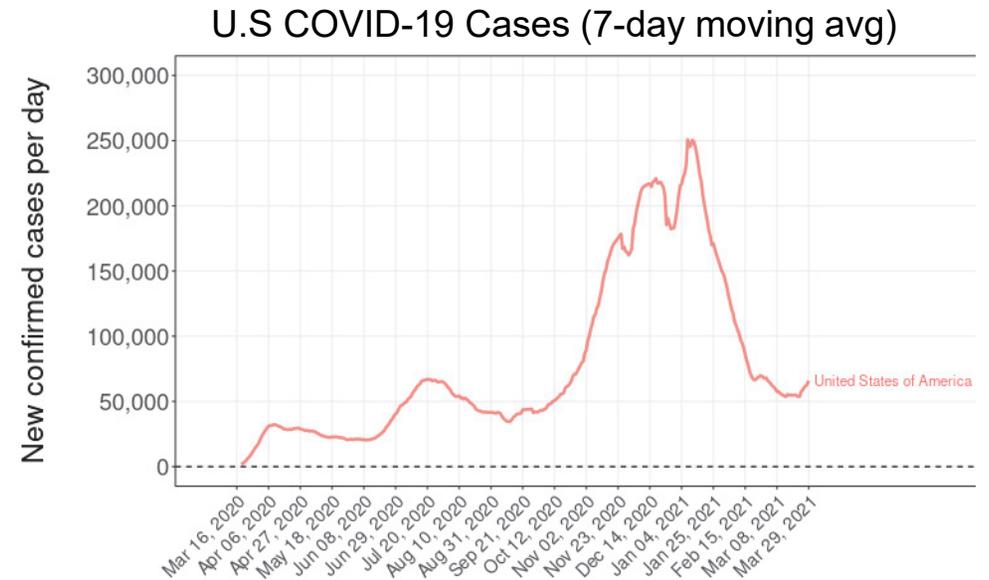
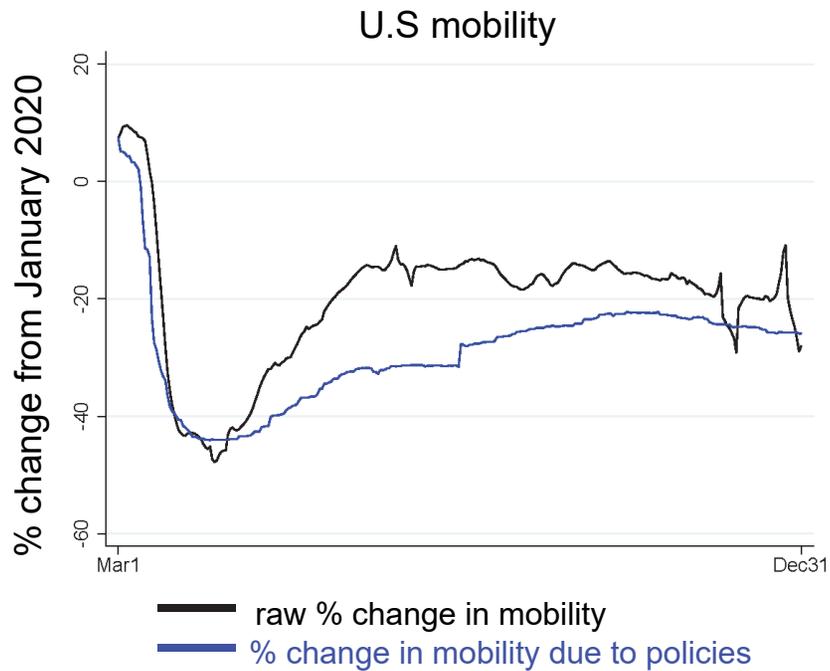
Mitigation policy stringency is also strongly correlated with reduced mobility to retail and recreation locations



— raw % change in mobility

— % change in mobility due to policies

How do reductions in mobility to retail and recreation locations affect COVID-19 case counts?



<https://test-science.shinyapps.io/covid-19-tracker/>

Mitigation policies that decreased traffic to retail and recreation locations decreased COVID-19 case counts in the U.S.

First stage results	U.S.	Avg Oxford Stringency Index Rating	Avg. percent change in retail & recreation mobility from policies	Avg Oxford Stringency Index Rating	Avg. percent change in retail & recreation mobility from policies
		March 2020 – July 2020		Aug 2020- Dec 2020	
		68.8	-25.2%	58.3	-15.4%

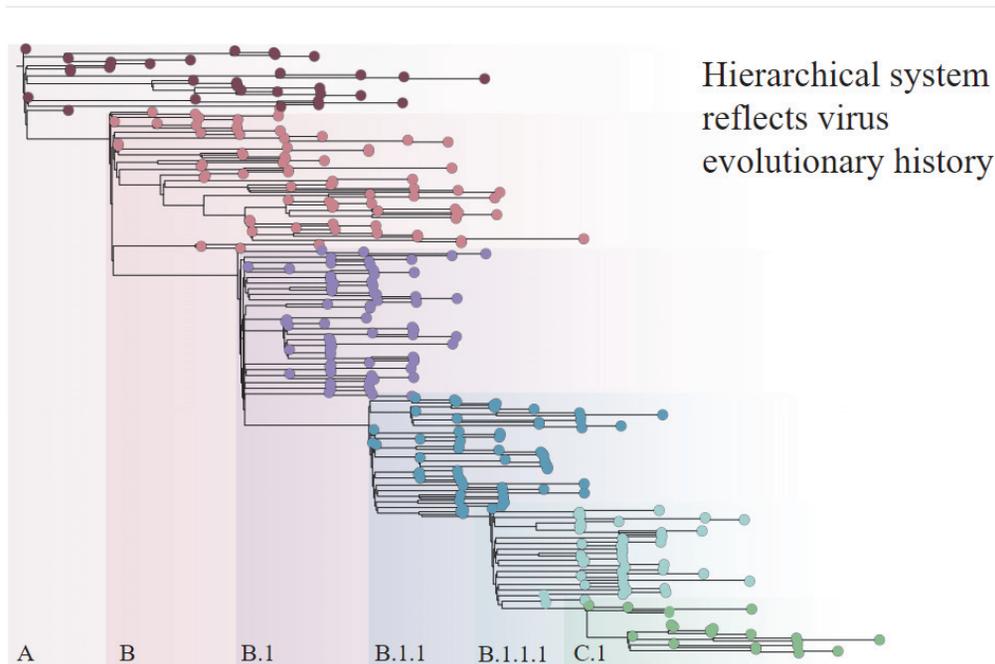
- Policies reduced traffic to retail and recreation locations by ~15-25%

Second stage results	U.S.	Avg. percent change in retail & recreation mobility from policies	Avg change in cases per 100k from policies	Avg. percent change in retail & recreation mobility from policies	Avg change in cases per 100k from policies
		March 2020 – July 2020		Aug 2020- Dec 2020	
		25.2%	-17.8	15.4%	-10.9

- The reduction in traffic to retail and recreation locations reduced COVID-19 case counts by ~10-17 cases per 100,000 per day

Genomic sequencing surveillance

Rapid, high-throughput genomic sequencing has resulted in unprecedented insight into how SARS-CoV-2 is evolving



*Note that C.1 is equivalent to B.1.1.1.1, with C being an alias for B.1.1.1 to shorten the name so it does not become infinitely long while still preserving the link to the parent lineages

PANGO (Rambaut et al. 2020 *Nat Microbiol*) is the algorithm behind the new lineages making the news.

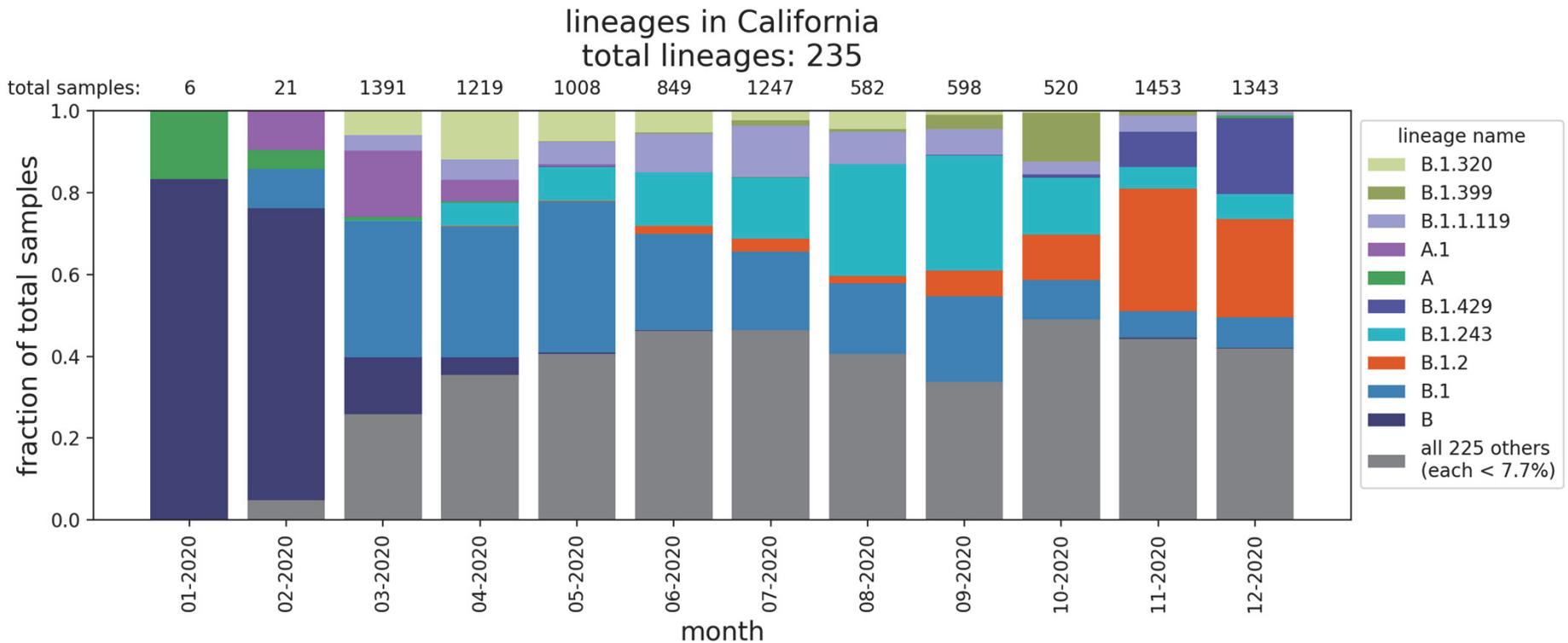
Lineages reflect:

- 1) A virus's ancestry
- 2) Geographic location
- 3) A new mutation (in that lineage)

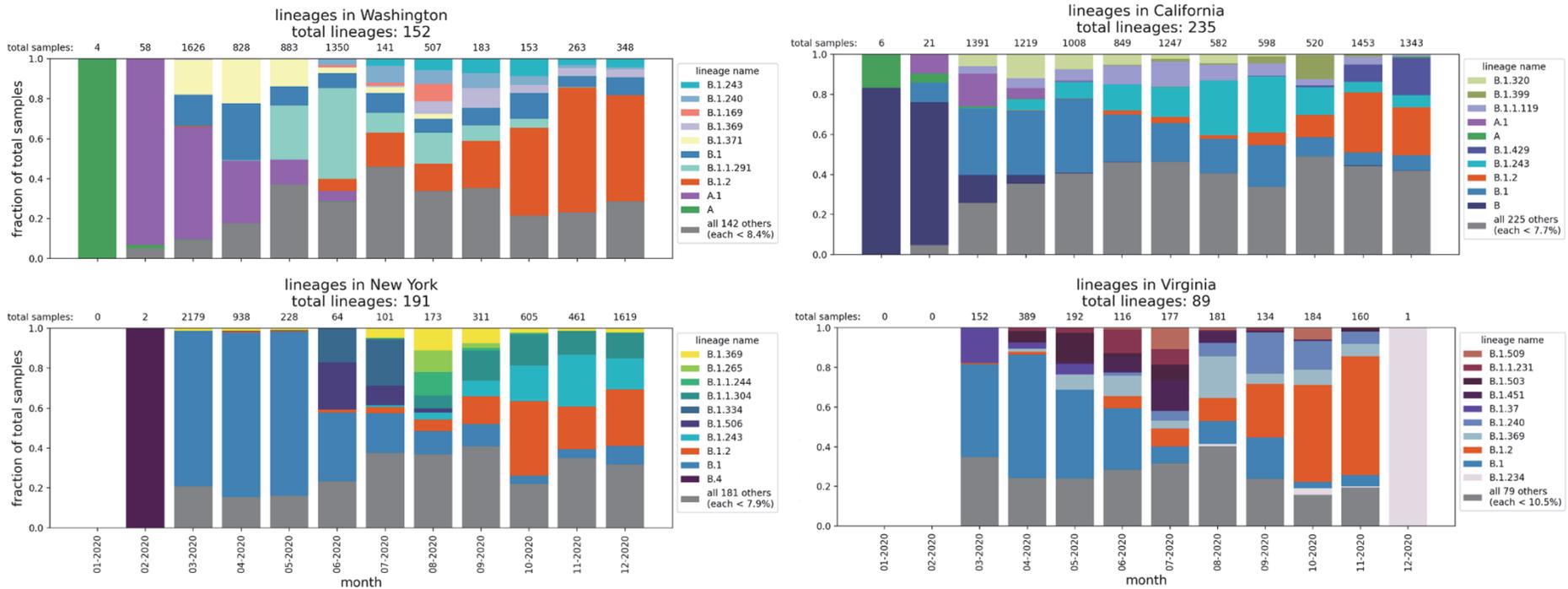
Lineages DO NOT reflect:

- 1) Whether a mutation is meaningful
- 2) Whether a mutation is unique
(the same mutation can occur in different lineages)

Using genomic sequence data, we can track lineages over time in the U.S.



SARS-CoV-2 lineages have ebbed and flowed over the course of the pandemic in the U.S.



- It is easy to determine what a virus’s genetic sequence is. It is more difficult to tell what the sequence means. **Lineage names reflect sequence, not meaning**
- Things that may cause a lineage to become predominant:
 - Founder effect – the reduction in genetic variation when a small subset of a large population is used to establish a new population
 - One or repeated superspreader events with that lineage
 - Small-number statistics/undersampling/biased sampling
 - A mutation with a biological mechanism that facilitates higher infectivity/spread

Thoughts on a pandemic in the time of “Big Data”

- More data is good!
 - Allows for real-time analysis to support deployment of resources and policy implementation
- Where more data or new data might complicate things:
 - Lack of insight into data collection issues, important caveats, issues that might affect analysis and interpretation
 - Example: Early in the pandemic, some states were reporting both PCR and serology test results into a rolled-up number, providing a false picture of real-time spread of the virus
 - Lack of experience with certain data types
 - Example: Genomic sequencing generating concern about emerging variants

Institute for Defense Analyses COVID-19 Team

Mr. Nathaniel Cleaves

Mr. Daniel Chaiken

Dr. Katherine Fisher

Dr. Kristen Guerrera

Mr. Ryan Murphy

Mr. Chris Oswald

Dr. Bryan Roberts

Dr. Zach Szlendak

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