

Network Systems
Science & Advanced
Computing

Biocomplexity Institute
& Initiative

University of Virginia

Estimation of COVID-19 Impact in Virginia

November 11th, 2020

(data current to November 10th)

Biocomplexity Institute Technical report: TR 2020-138



BIOCOMPLEXITY INSTITUTE

biocomplexity.virginia.edu

About Us

- Biocomplexity Institute at the University of Virginia
 - Using big data and simulations to understand massively interactive systems and solve societal problems
- Over 20 years of crafting and analyzing infectious disease models
 - Pandemic response for Influenza, Ebola, Zika, and others



Points of Contact

Bryan Lewis
brylew@virginia.edu

Srini Venkatramanan
srini@virginia.edu

Madhav Marathe
marathe@virginia.edu

Chris Barrett
ChrisBarrett@virginia.edu

Biocomplexity COVID-19 Response Team

Aniruddha Adiga, Abhijin Adiga, Hannah Baek, Chris Barrett, Golda Barrow, Richard Beckman, Parantapa Bhattacharya, Andrei Bura, Jiangzhuo Chen, Clark Cucinell, Patrick Corbett, Allan Dickerman, Stephen Eubank, Arindam Fadikar, Joshua Goldstein, Stefan Hoops, Ben Hurt, Sallie Keller, Ron Kenyon, Brian Klahn, Gizem Korkmaz, Vicki Lancaster, Bryan Lewis, Dustin Machi, Chunhong Mao, Achla Marathe, Madhav Marathe, Fanchao Meng, Henning Mortveit, Mark Orr, Joseph Outten, Akhil Peddireddy, Przemyslaw Porebski, SS Ravi, Erin Raymond, Jose Bayoan Santiago Calderon, James Schlitt, Aaron Schroeder, Stephanie Shipp, Samarth Swarup, Alex Telionis, Srinivasan Venkatramanan, Anil Vullikanti, James Walke, Amanda Wilson, Dawen Xie



Overview

- **Goal:** Understand impact of COVID-19 mitigations in Virginia
- **Approach:**
 - Calibrate explanatory mechanistic model to observed cases
 - Project infections through December
 - Consider a range of possible mitigation effects in "what-if" scenarios
- **Outcomes:**
 - Ill, Confirmed, Hospitalized, ICU, Ventilated, Death
 - Geographic spread over time, case counts, healthcare burdens

Key Takeaways

Projecting future cases precisely is impossible and unnecessary.

Even without perfect projections, we can confidently draw conclusions:

- **Virginia continues steady growth recording highest case rates of epidemic**
- VA mean weekly incidence (16.8/100K) is up again (from 14.8) though slower than nationally (46/100K from 34/100K).
- Projections are mostly up, showing potential for strain on health care system in some regions as early as December.
- Recent updates:
 - Horizon extended to March 1st
 - Preliminary results of ensemble of forecasting models included
 - Planning scenarios and case ascertainment rates remain as updated in previous weeks
- The situation is changing rapidly. Models will be updated regularly.

Situation Assessment



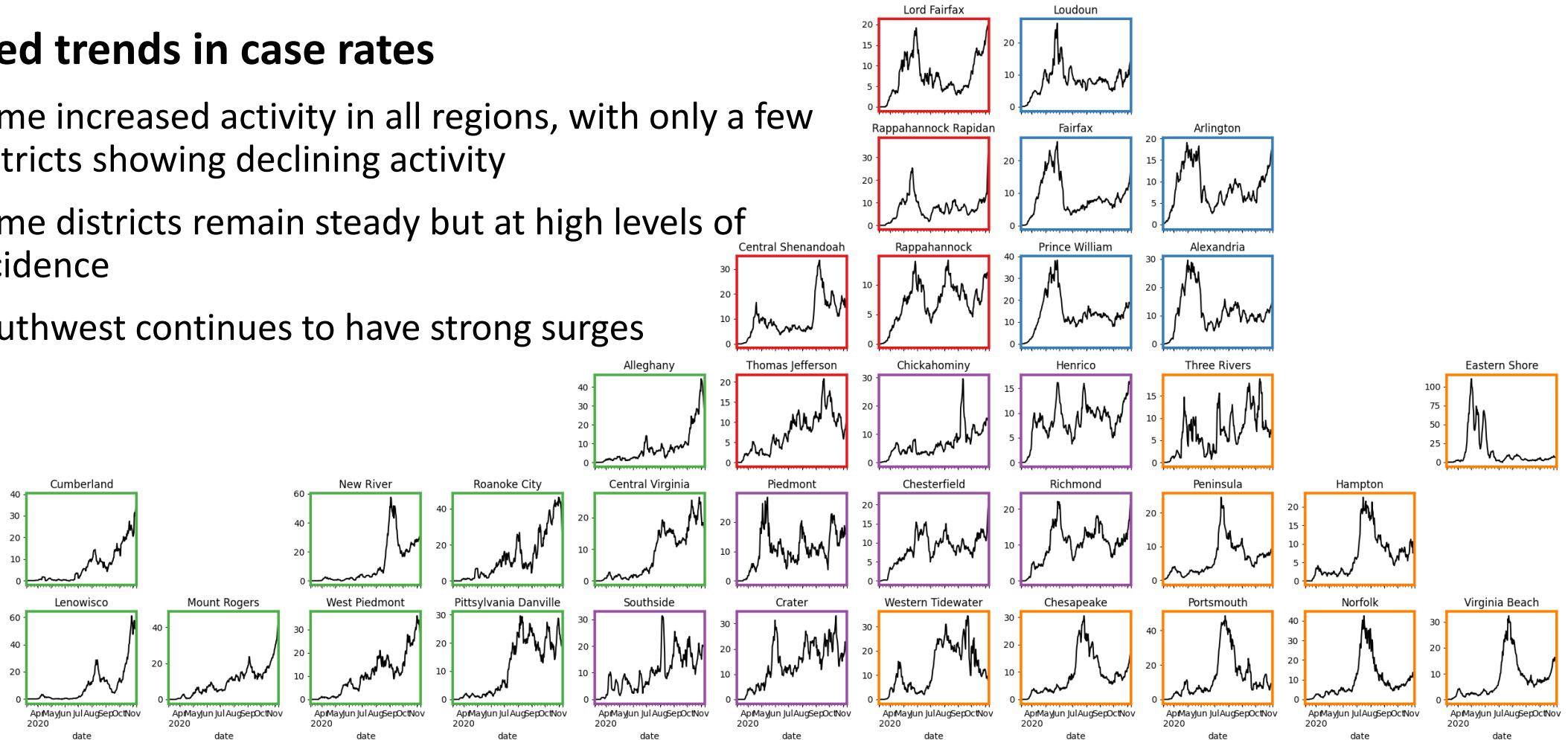
UNIVERSITY OF VIRGINIA

BIOCOMPLEXITY INSTITUTE

Case Rate (per 100k) by VDH District

Mixed trends in case rates

- Some increased activity in all regions, with only a few districts showing declining activity
- Some districts remain steady but at high levels of incidence
- Southwest continues to have strong surges

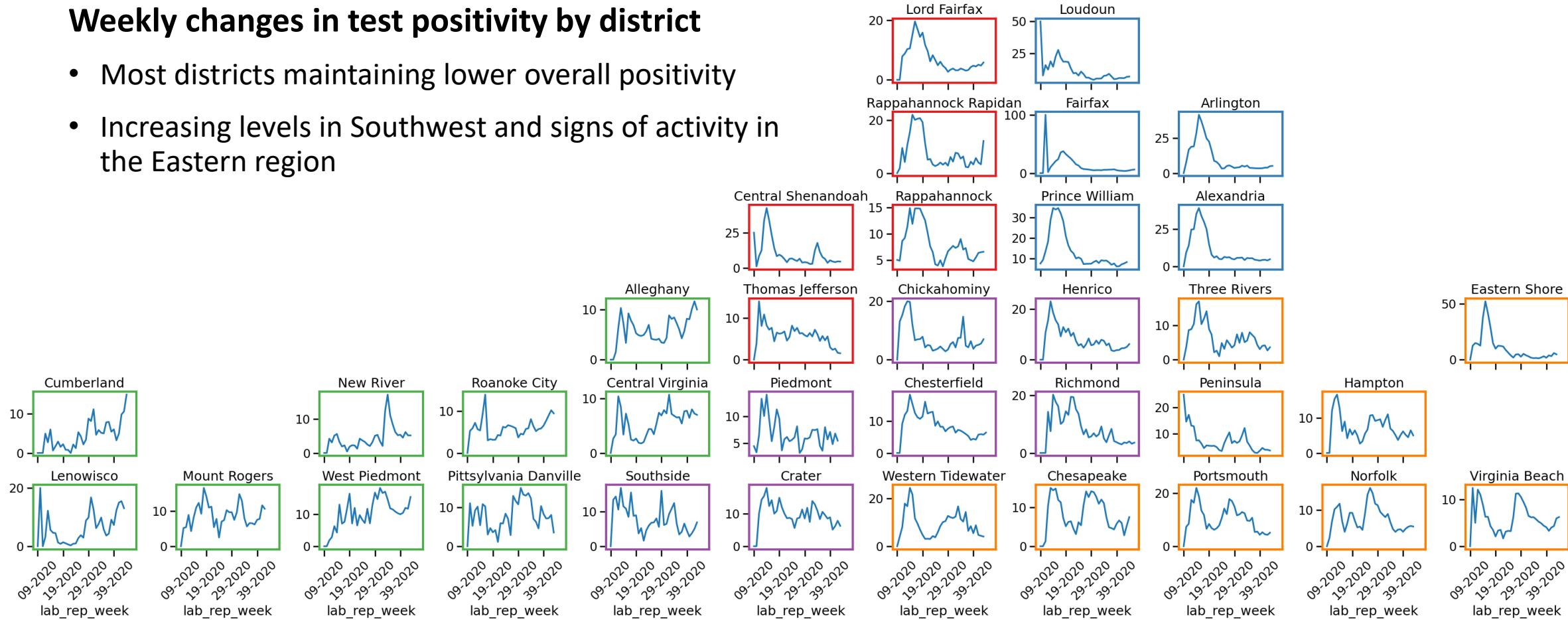


UNIVERSITY OF VIRGINIA

Test Positivity by VDH District

Weekly changes in test positivity by district

- Most districts maintaining lower overall positivity
- Increasing levels in Southwest and signs of activity in the Eastern region



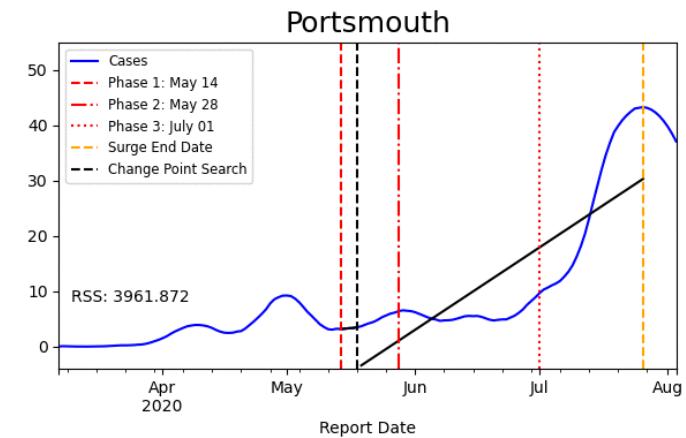
UNIVERSITY OF VIRGINIA

District Trajectories

Goal: Define epochs of a Health District's COVID-19 incidence to characterize the current trajectory

Method: Find recent peak and use hockey stick fit to find inflection point afterwards, then use this period's slope to define the trajectory

Hockey stick fit



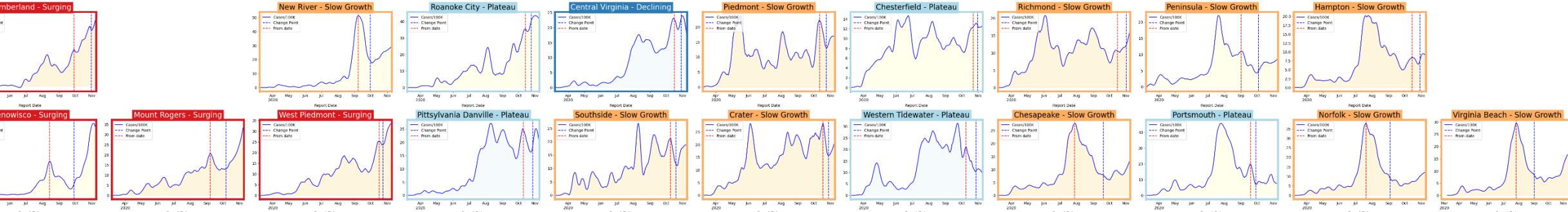
Trajectory	Description	Weekly Case Rate (per 100K) bounds	# Districts (last week)
Declining	Sustained decreases following a recent peak	below -0.9	2 (5)
Plateau	Steady level with minimal trend up or down	above -0.9 and below 0.5	8 (10)
Slow Growth	Sustained growth not rapid enough to be considered a Surge	above 0.5 and below 2.5	17 (17)
In Surge	Currently experiencing sustained rapid and significant growth	2.5 or greater	8 (3)



District Trajectories

Status	# Districts (last week)
Declining	2 (5)
Plateau	8 (10)
Slow Growth	17 (17)
In Surge	8 (3)

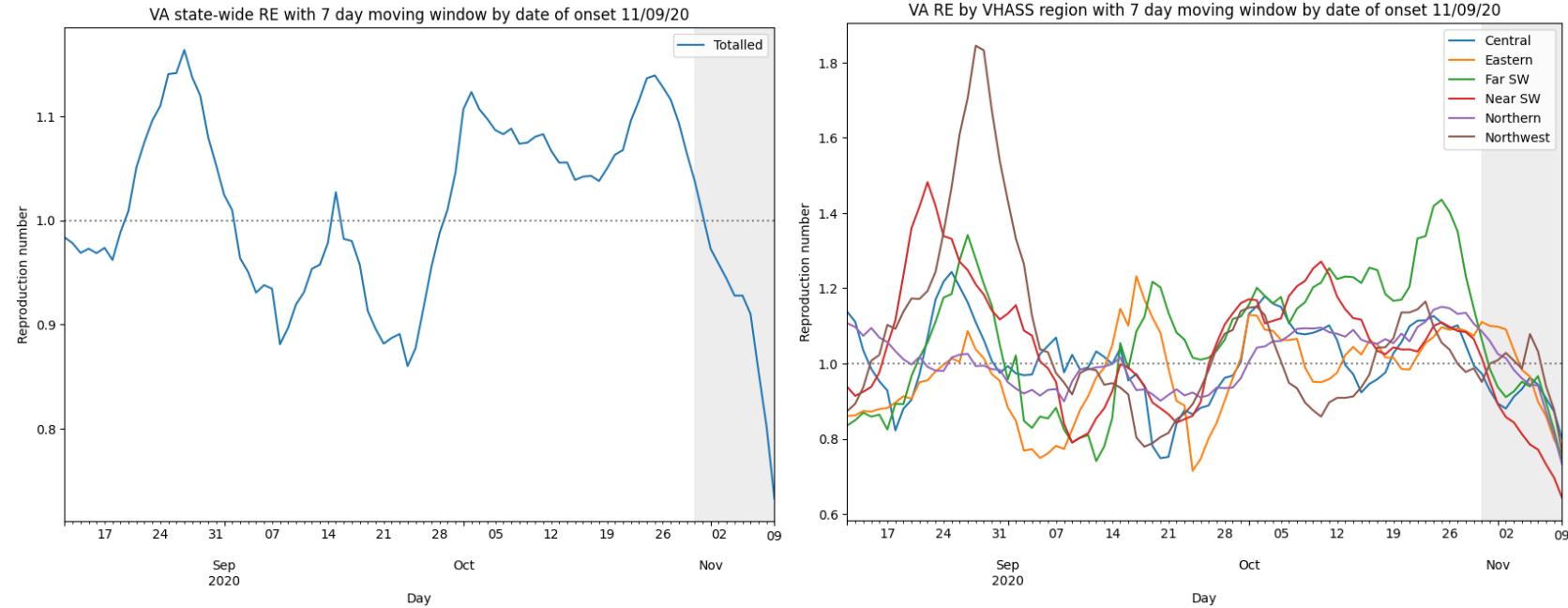
Curve shows smoothed case rate (per 100K)
 Trajectories of states in label & chart box
 Case Rate curve colored by Reproductive



Estimating Daily Reproductive Number

October 31st Estimates

Region	Current R_e	Diff Last Week
State-wide	1.005	-0.082
Central	0.929	-0.151
Eastern	1.100	0.054
Far SW	0.986	-0.380
Near SW	0.952	-0.072
Northern	1.060	-0.032
Northwest	1.000	-0.074



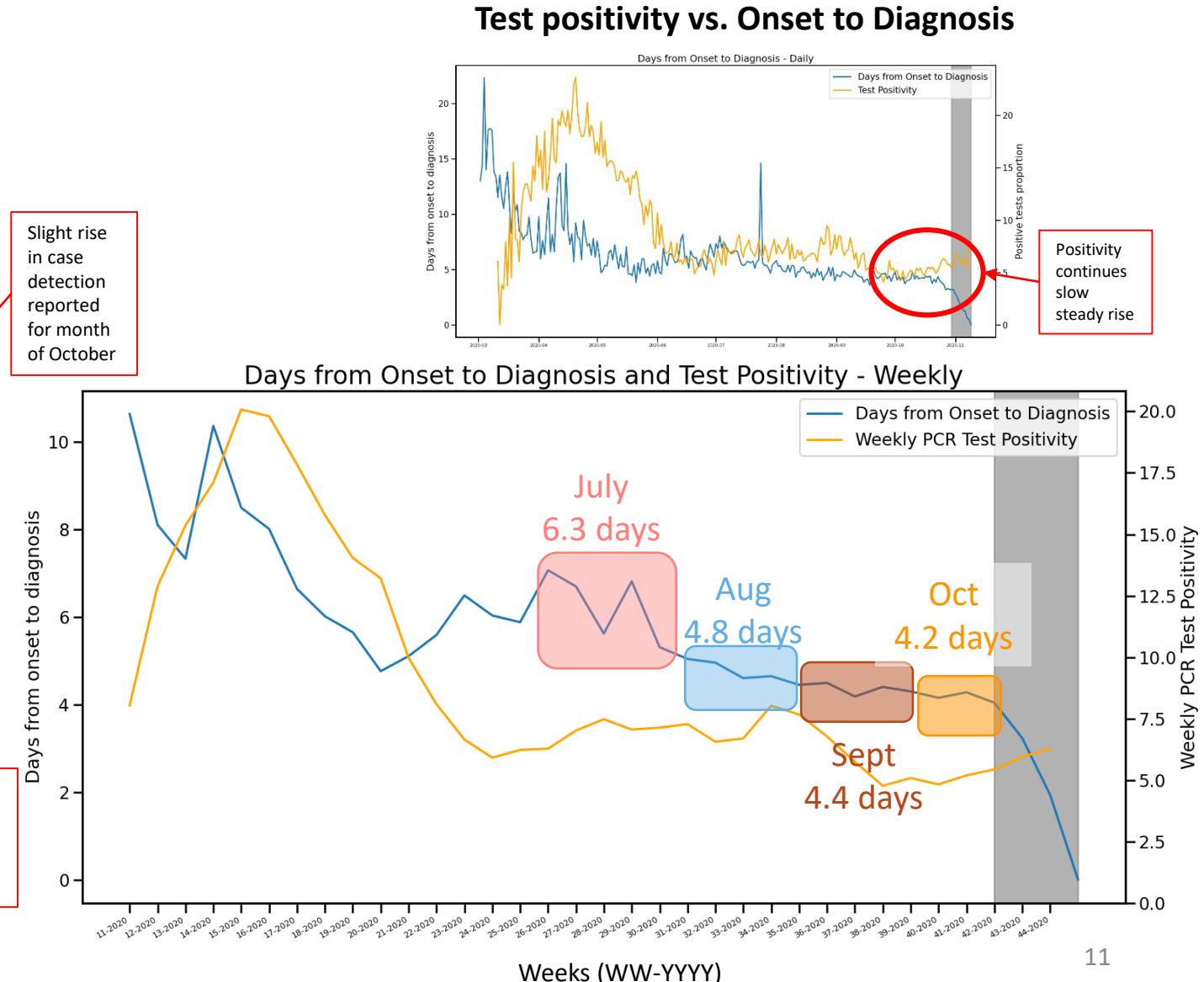
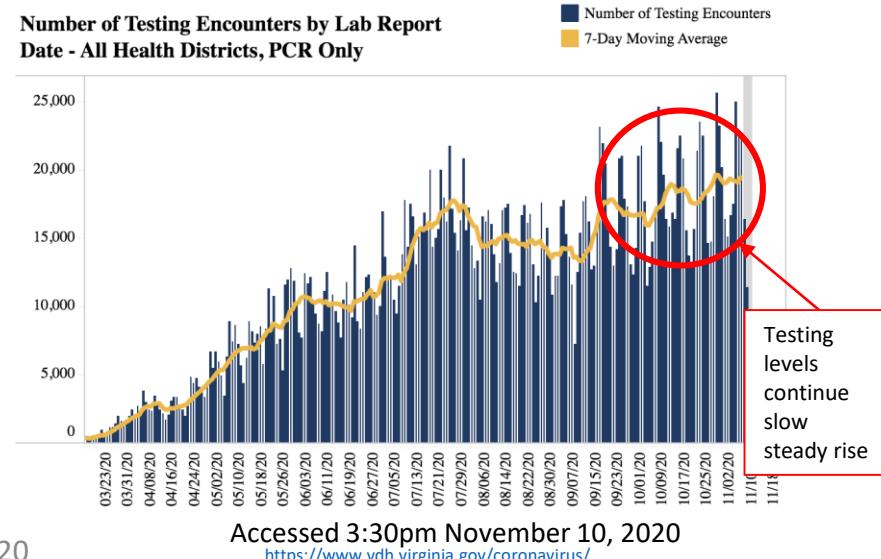
Methodology

- Wallinga-Teunis method (EpiEstim¹) for cases by date of onset
- Serial interval: 6 days (2 day std dev)
- Recent estimates may be unstable due to backfill

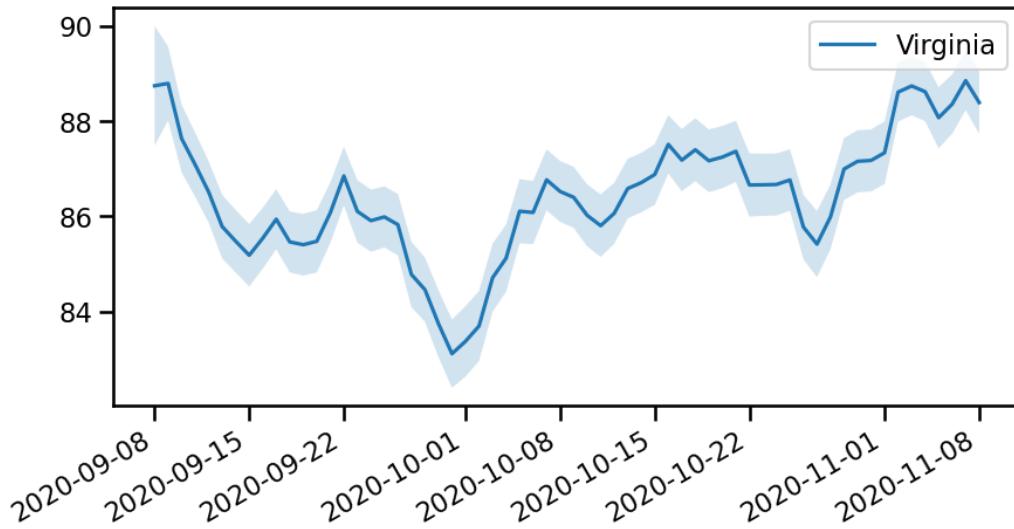
1. Anne Cori, Neil M. Ferguson, Christophe Fraser, Simon Cauchemez. A New Framework and Software to Estimate Time-Varying Reproduction Numbers During Epidemics. American Journal of Epidemiology, Volume 178, Issue 9, 1 November 2013, Pages 1505–1512, <https://doi.org/10.1093/aje/kwt133>

Changes in Case Detection

Timeframe (weeks)	Mean days	% difference from overall mean
April (13-16)	8.6	50%
May (17-21)	5.6	-1%
June (22-25)	6.0	5%
July (26-30)	6.3	10%
Aug (31-34)	4.8	-16%
Sept (35-38)	4.4	-23%
Oct (39-42)	4.2	-27%
Overall (13-42)	5.7	0%



Mask usage in Virginia

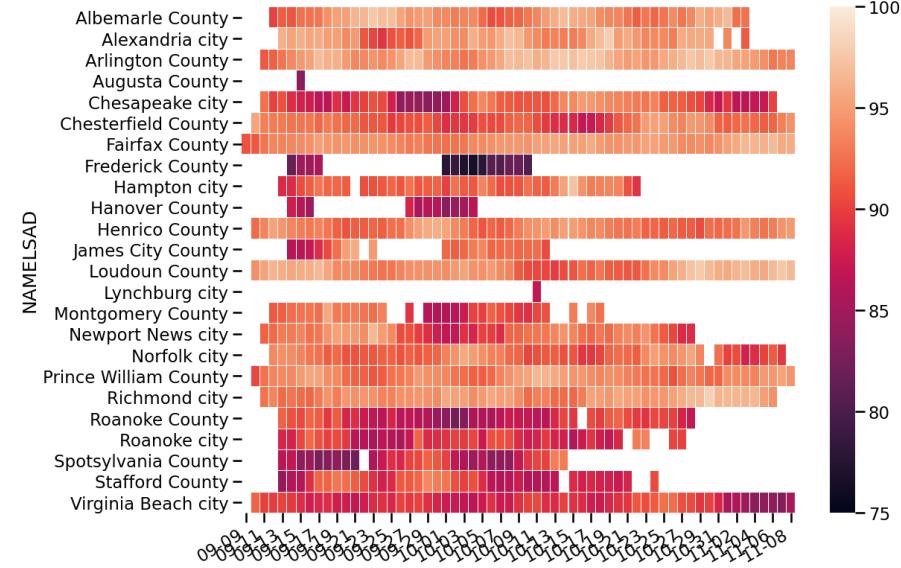


State level mask usage as reported via Facebook surveys over the past month shows ranges from 83% to 89%

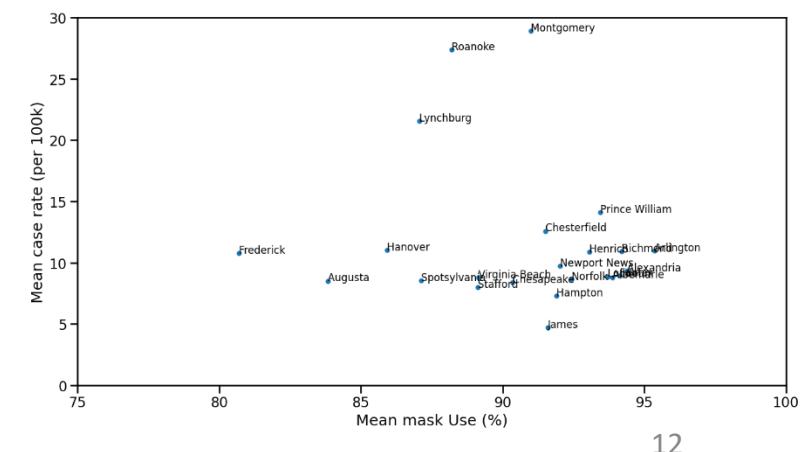
- Relatively stable over time
- Limited variance across the commonwealth
- ~3000 daily responses from VA

Data Source: <https://covidcast.cmu.edu>

Correlations seen at national level with mask use and case rate not emerging across VA counties, reflecting high use across commonwealth



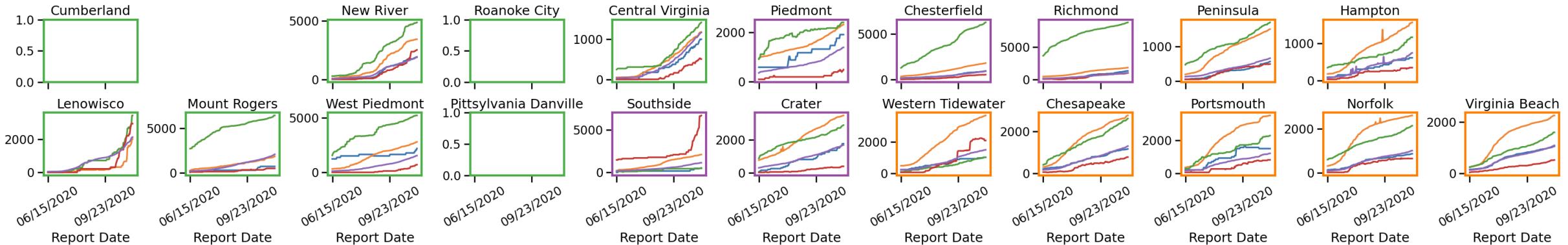
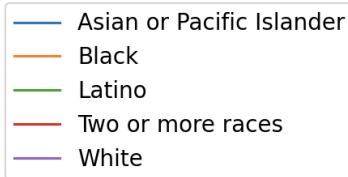
Some county level fluctuations since beginning of Sept., though data quality may be affected by sample sizes.



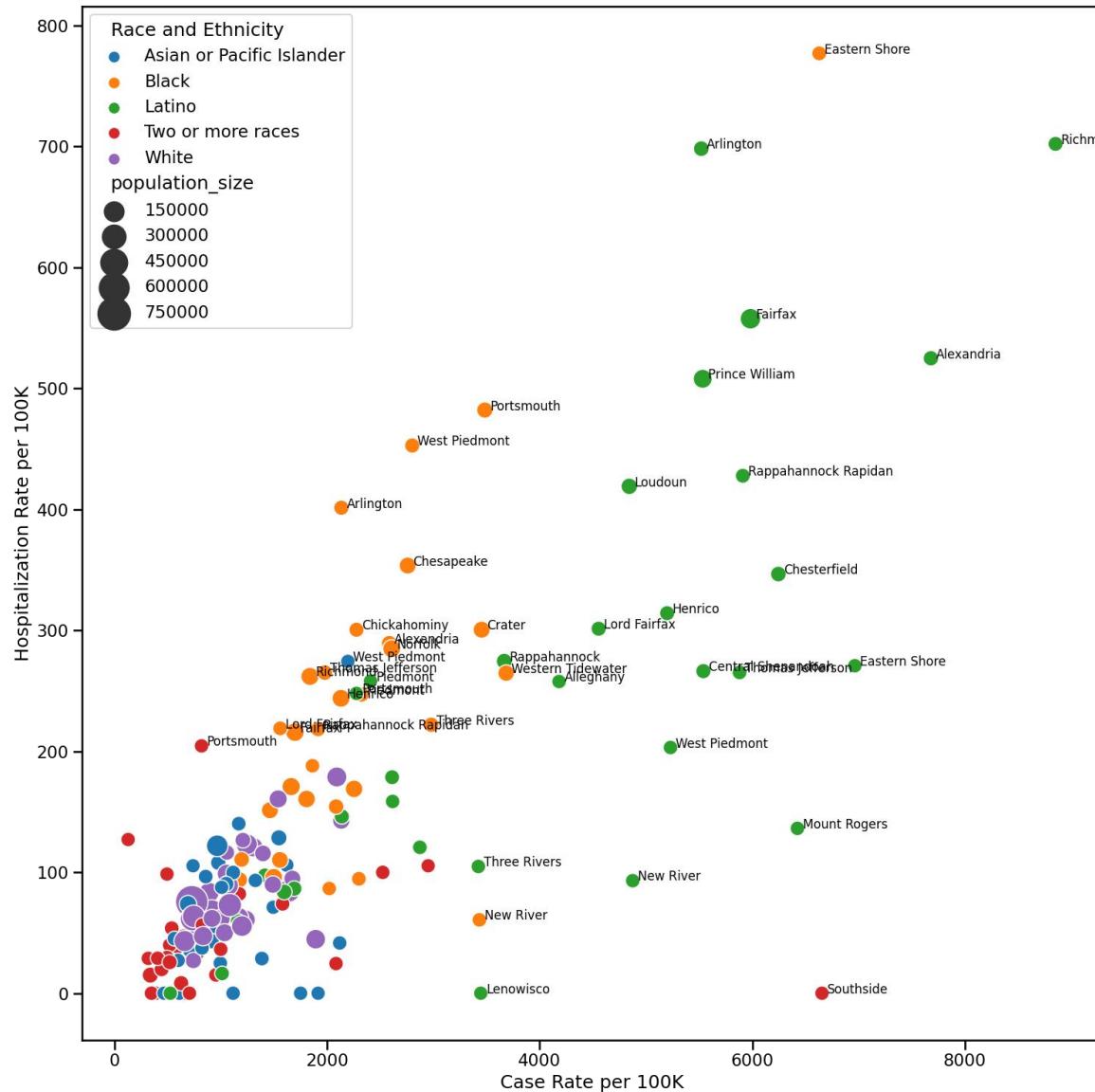
Race and Ethnicity Attack Rates (per 100K)

Cumulative Race and Ethnicity Attack Rates (per 100k)

- Black and Latinx populations have much higher case, hospitalization, and death rates
- Disparity is more pronounced in some districts than others
- Based on 2019 census race-ethnicity data by county



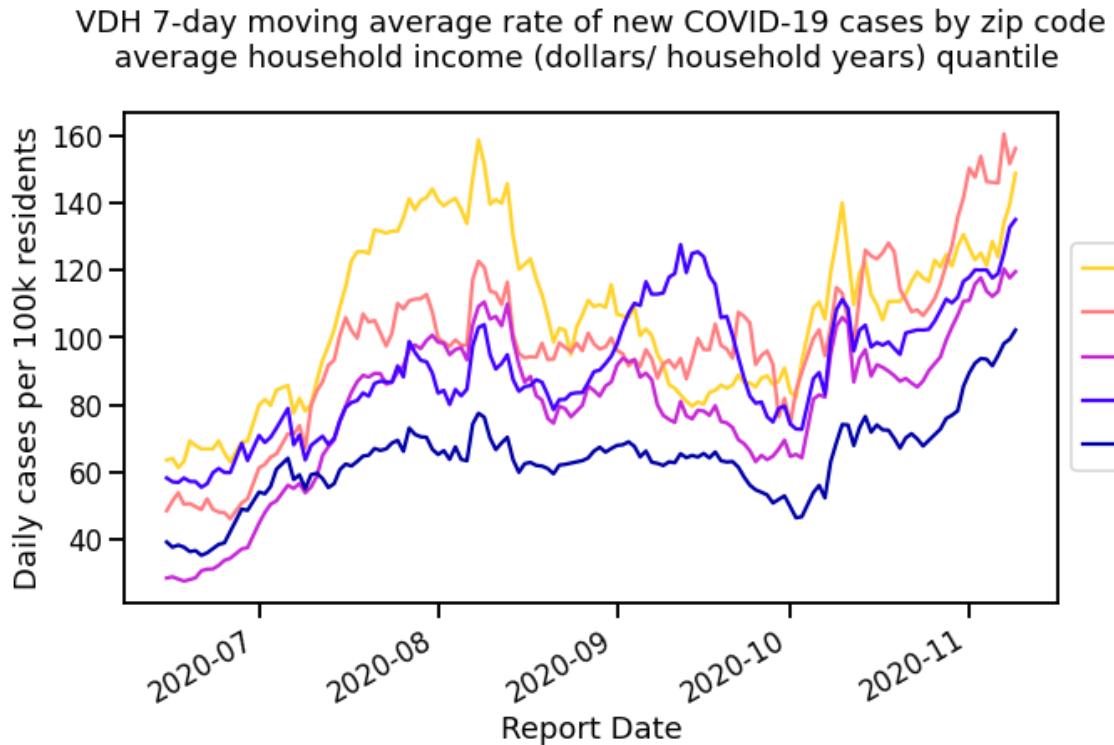
Race and Ethnicity cases per 100K



Rates per 100K of each Racial-Ethnic population by Health District

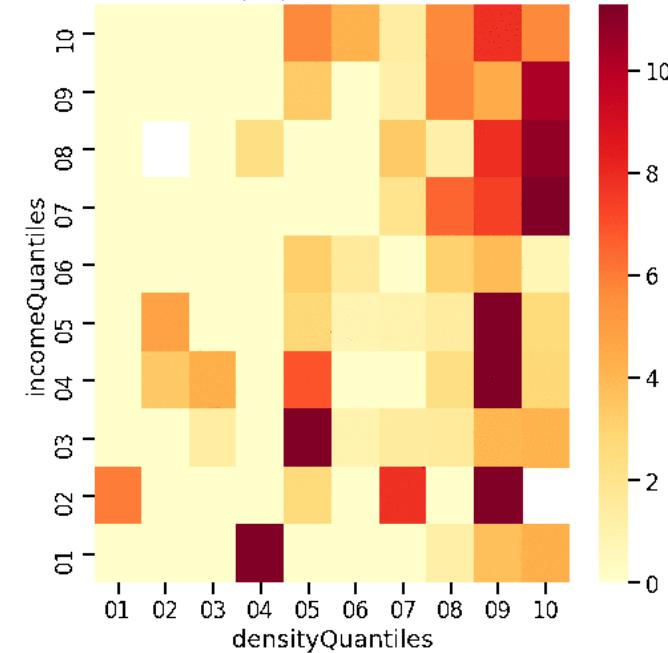
- Each Health District's Racial-Ethnic population is plotted by their Hospitalization and Case Rate
- Points are sized based on their overall population size
- Overlapping labels removed for clarity

Impact across Density and Income



All zipcodes show steady growth, with lowest incomes showing the most rapid upticks

VDH mean cases per 100k by zip code population density (person/ sq mile)
and average household income (dollars/ household years) quantiles
06/03/20 - 06/09/20



Full evolution of pandemic, shows shifts from denser and wealthier zip codes to poorer and less dense zip codes, followed by a repeat of the pattern. Recently see an uptick across the spectrum of density and income

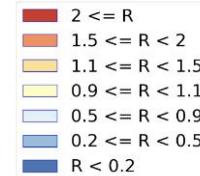
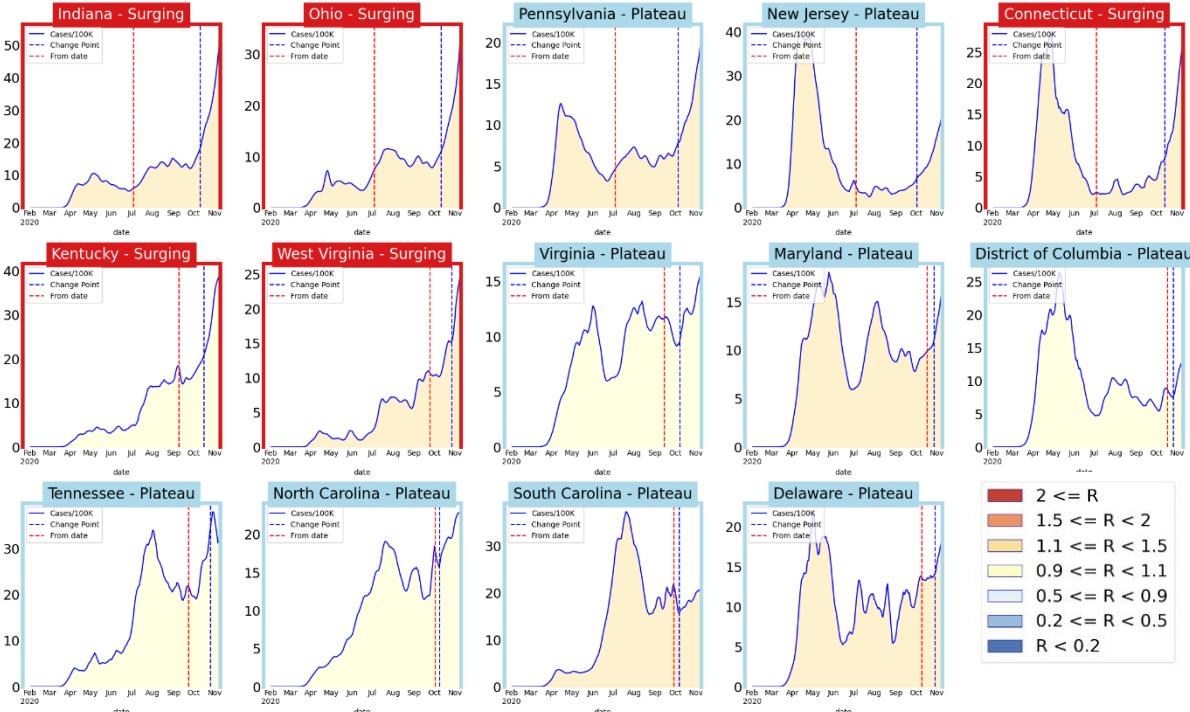


UNIVERSITY OF VIRGINIA

BIOCOMPLEXITY INSTITUTE

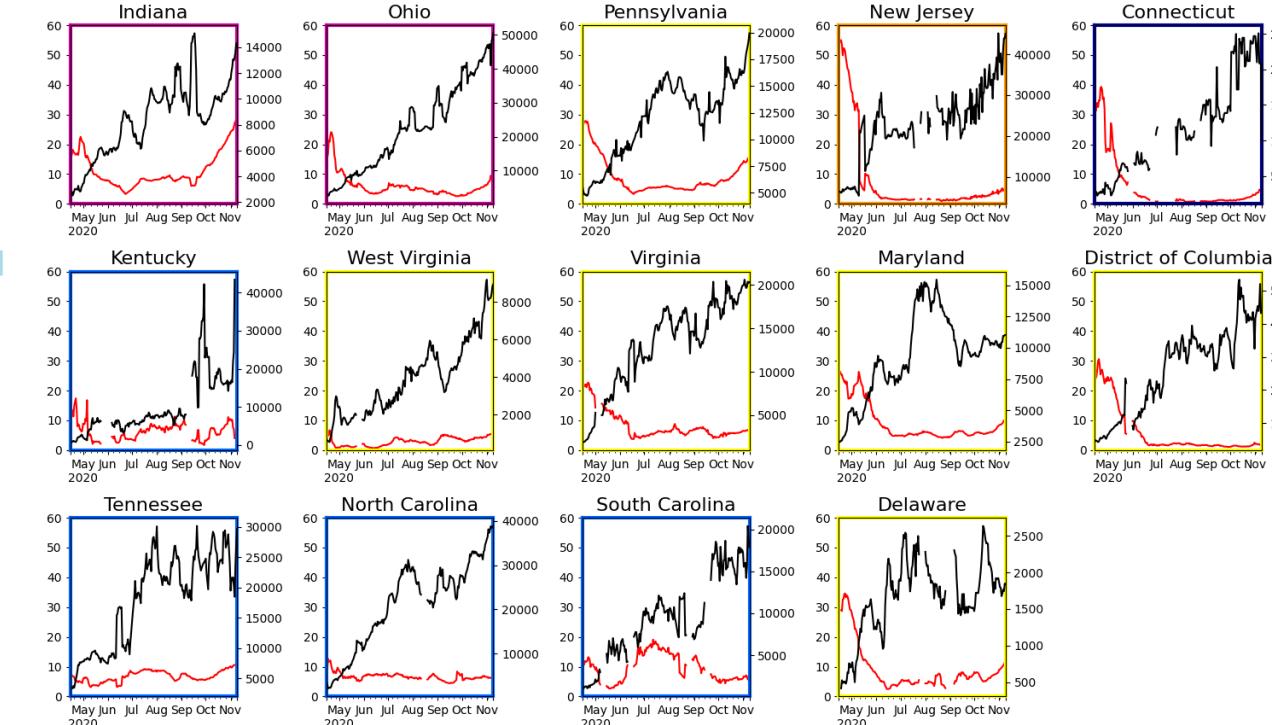
Other State Comparisons

Trajectories of States



- VA and most other mid-Atlantic states have upward trends KY, WV along with OH, IN, and CT remain in surge (24 total in US)
- TN shows some signs of slowing

Tests per Day and Test Positivity

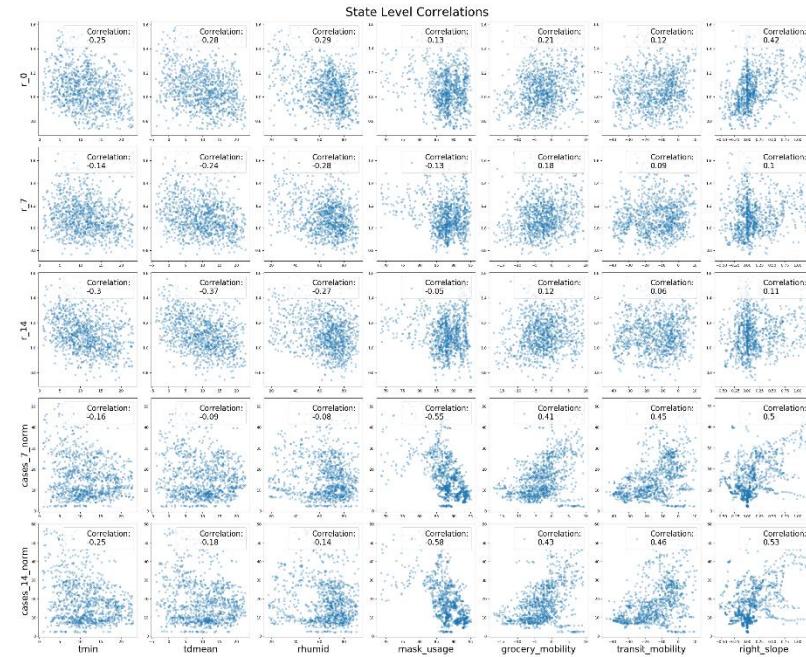
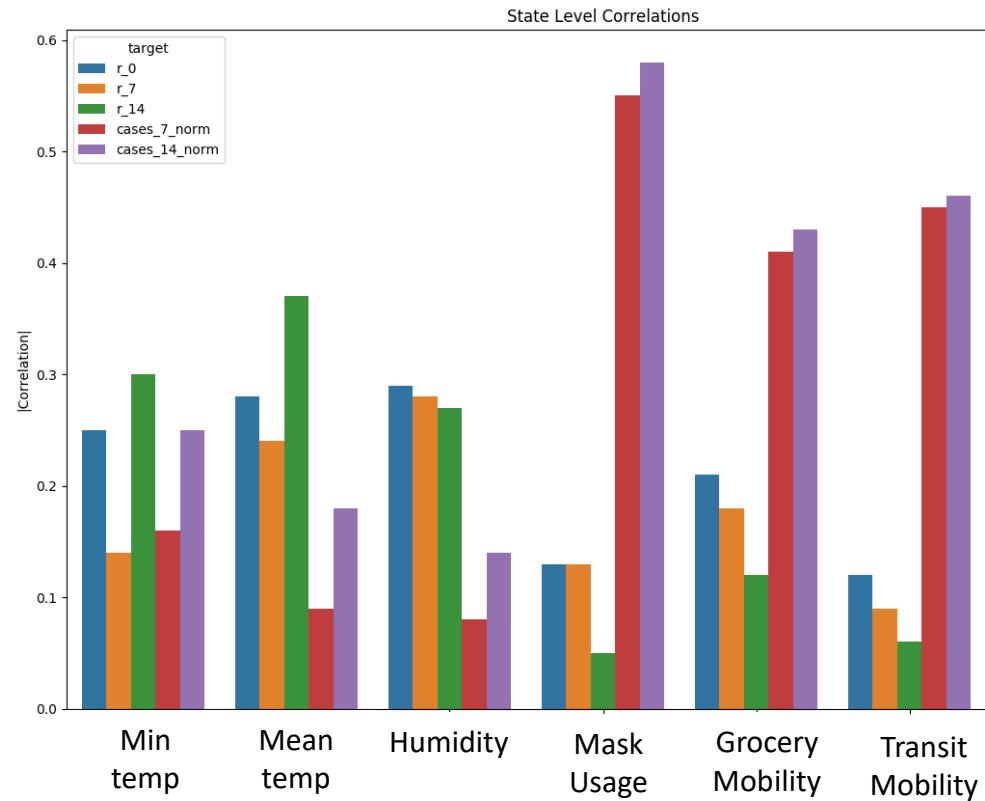
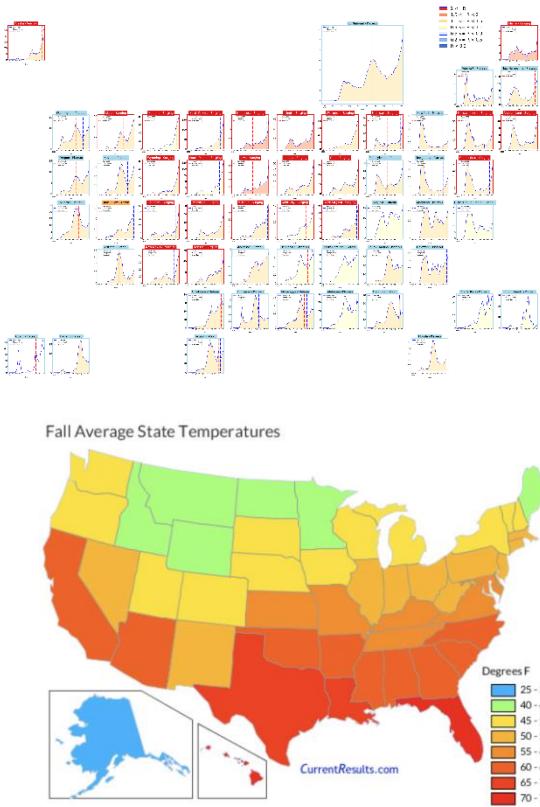


- Test positivity mixed, VA's rate has started slowly growing
- Testing volumes remain steady and relatively high in most states



UNIVERSITY OF VIRGINIA

Growth Associated with Temperature and Humidity



- As weather cools and humidity drops, COVID-19 survival and chance of transmission may rise
- Correlations with other factors are also strong for R (0, 7, 14 day delay) and confirmed cases (7 and 14 day delay)
- Weather variables better correlation with R estimates, while mobility and mask usage correlate well with case rates

Zip code level weekly Case Rate (per 100K)

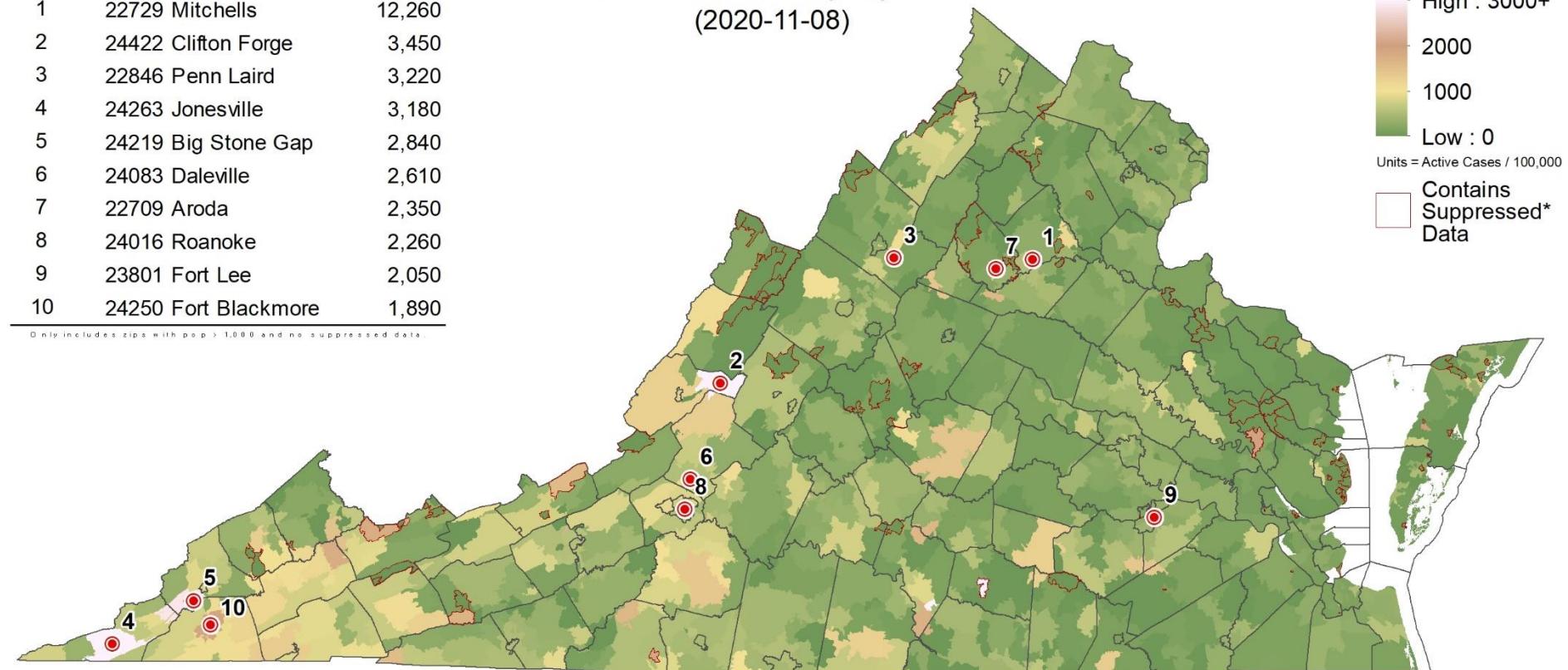
Case Rates in the last week by zip code

- Concentrations of very high prevalence in some zip codes
- Trend back towards very high rates in a few zips and lower in surrounding areas
- Southwest has considerable concentration of high prevalence zips
- Some counts are low and suppressed to protect anonymity, those are shown in white

Rank	Zip Code Name	Prevalence
1	22729 Mitchells	12,260
2	24422 Clifton Forge	3,450
3	22846 Penn Laird	3,220
4	24263 Jonesville	3,180
5	24219 Big Stone Gap	2,840
6	24083 Daleville	2,610
7	22709 Aroda	2,350
8	24016 Roanoke	2,260
9	23801 Fort Lee	2,050
10	24250 Fort Blackmore	1,890

Only includes zips with pop > 1,000 and no suppressed data.

Point Prevalence by Zip Code
(2020-11-08)

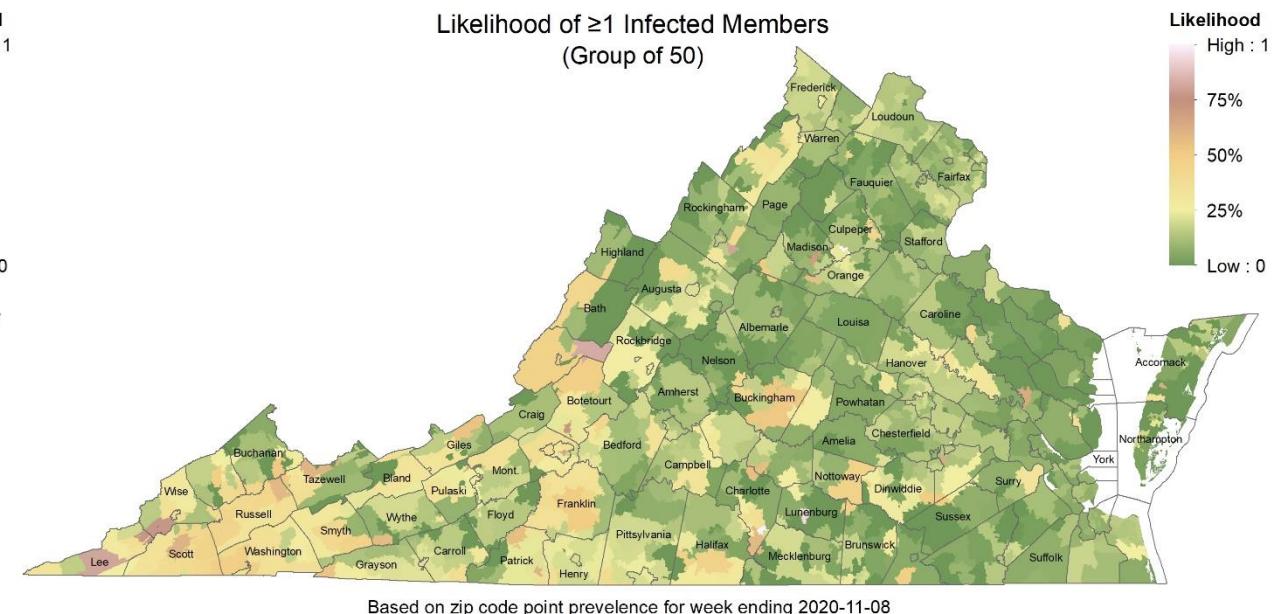
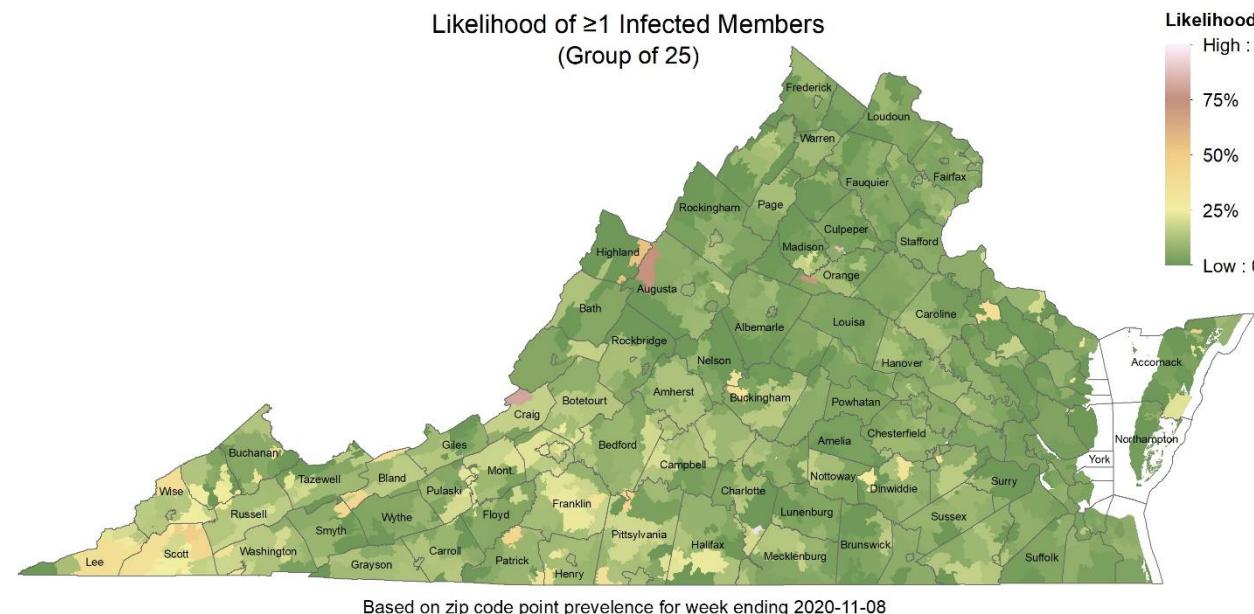


Point Prevalence
High : 3000+
2000
1000
Low : 0
Units = Active Cases / 100,000
Contains Suppressed* Data

Risk of Exposure by Group Size

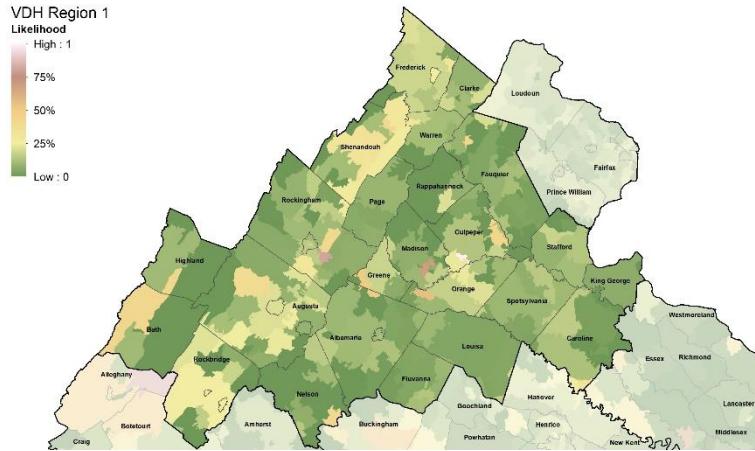
Case Prevalence in the last week by zip code used to calculate risk of encountering someone infected in a gathering of randomly selected people (group size 25 or 50)

- Assumes 3 undetected infections per confirmed case (ascertainment rate from recent seroprevalence survey)
- Moderate risk for groups of 50 across the commonwealth, especially in the southern half of the state
- Some zip codes have high likelihood of exposure even in groups of 25

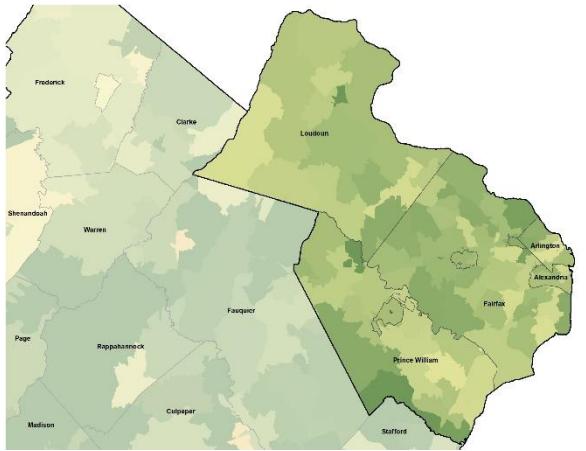


Zip code level weekly Case Rate (per 100K)

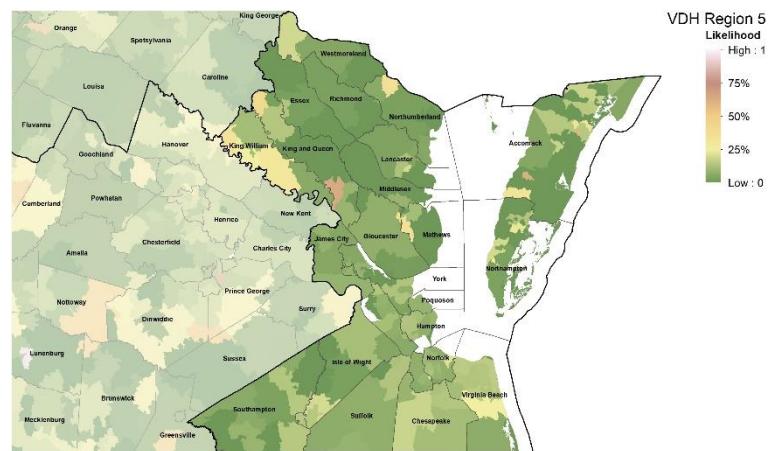
Northwest



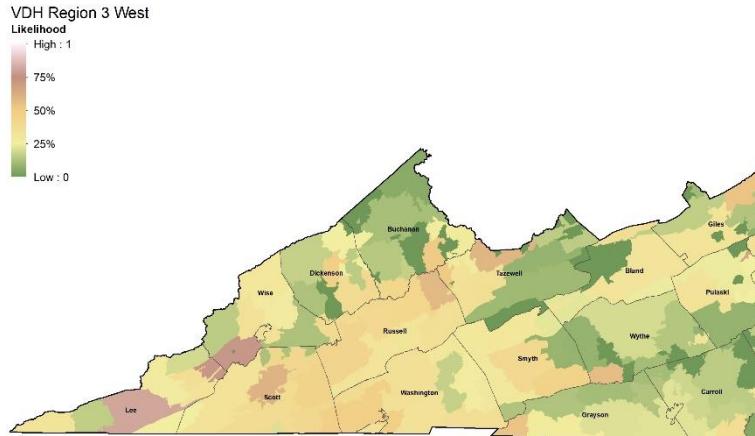
North



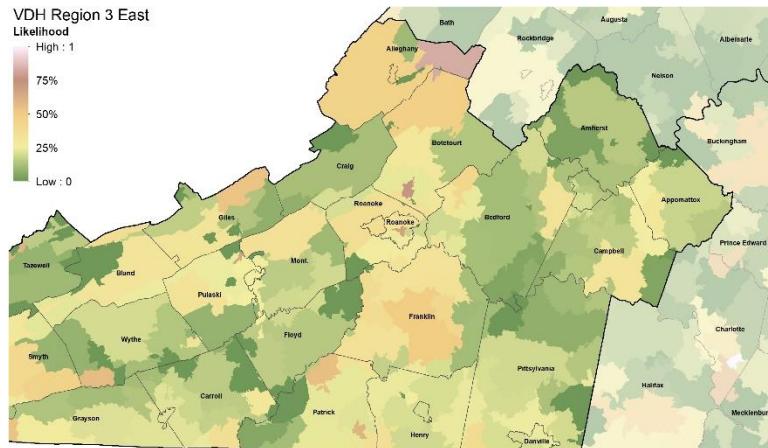
Eastern



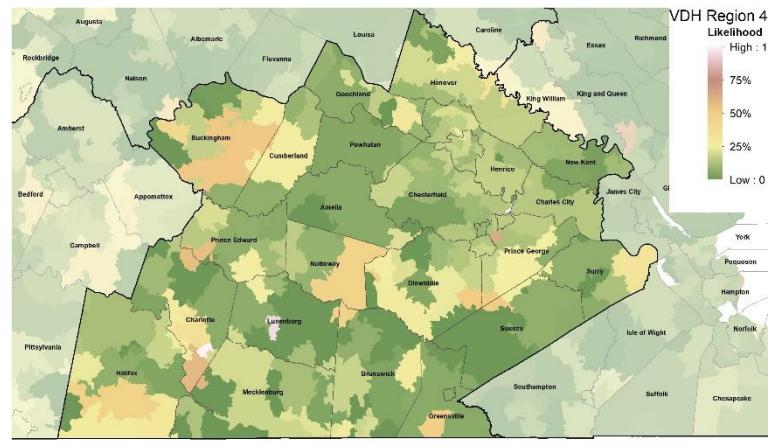
Far Southwest



Near Southwest

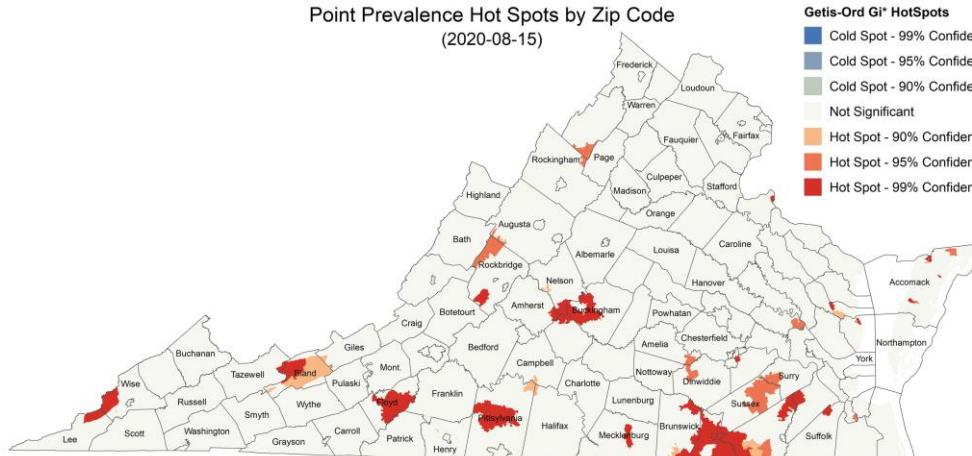


Central



Zip Code Hot Spots

Previous weeks



Hot Spot Significance	# of Zips (last week)
99%	7 (5)
95%	1 (4)
90%	2 (1)

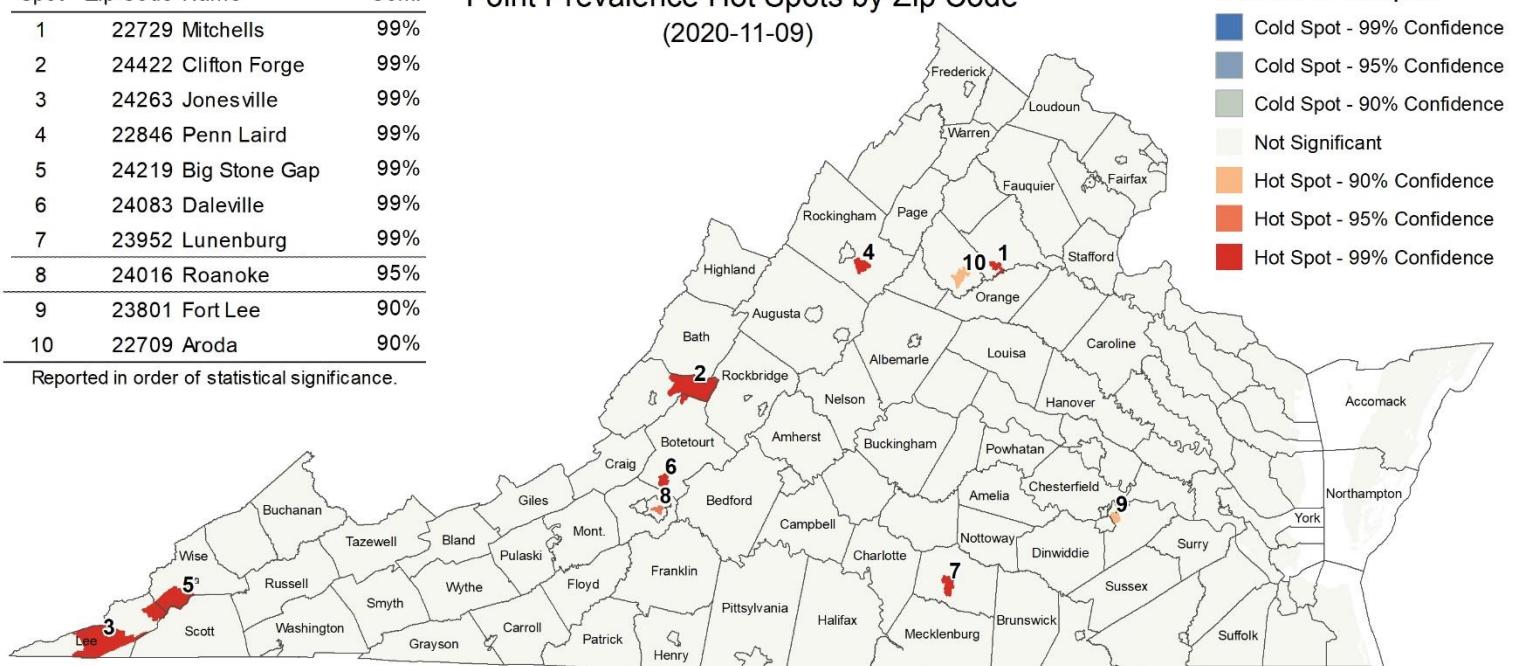
Hotspots across commonwealth

- Concentrated in the Far Southwest
- Captures some very high prevalence rates in some zips

Spot	Zip Code Name	Conf.
1	22729 Mitchells	99%
2	24422 Clifton Forge	99%
3	24263 Jonesville	99%
4	22846 Penn Laird	99%
5	24219 Big Stone Gap	99%
6	24083 Daleville	99%
7	23952 Lunenburg	99%
8	24016 Roanoke	95%
9	23801 Fort Lee	90%
10	22709 Aroda	90%

Reported in order of statistical significance.

Point Prevalence Hot Spots by Zip Code
(2020-11-09)



Model Update – Adaptive Fitting

Adaptive Fitting Approach

Each county fit precisely, with recent trends used for future projection

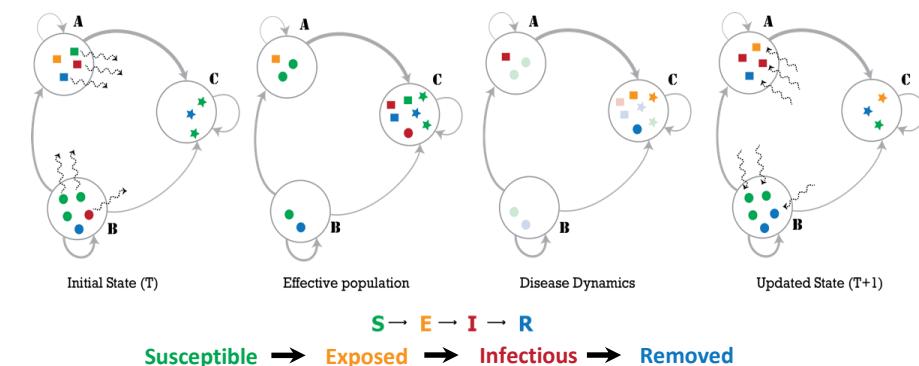
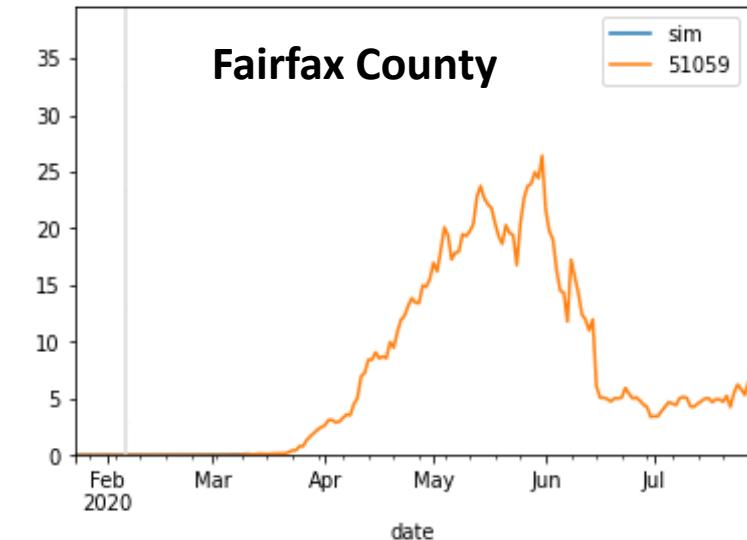
- Allows history to be precisely captured, and used to guide bounds on projections

Model: An alternative use of the same meta-population model, PatchSim

- Allows for future “what-if” Scenarios to be layered on top of calibrated model
- Eliminates connectivity between patches, to allow calibration to capture the increasingly unsynchronized epidemic

External Seeding: Steady low-level importation

- Widespread pandemic eliminates sensitivity to initial conditions
- Uses steady 1 case per 10M population per day external seeding



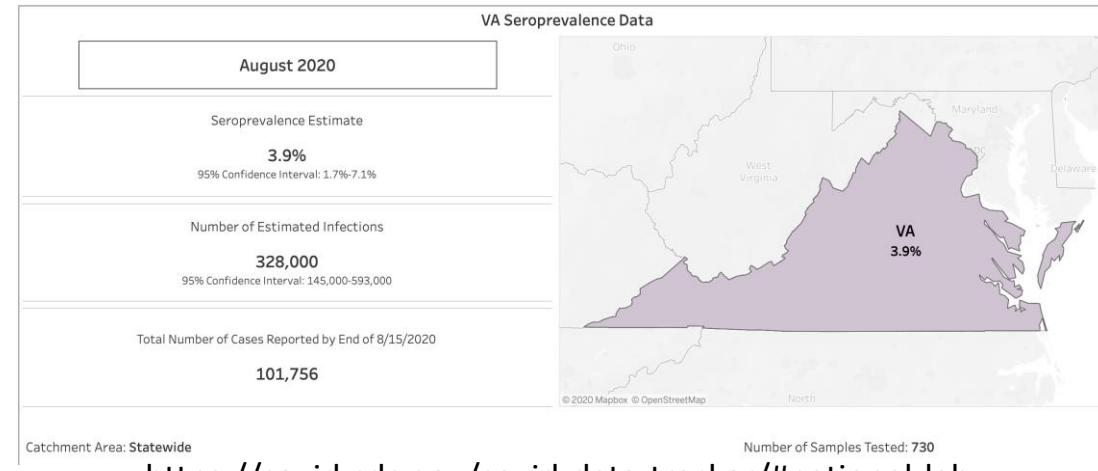
Seroprevalence updates to model design

Several seroprevalence studies provide better picture of how many actual infections have occurred

- Virginia Serology Study estimated 2.4% of Virginians estimated infected (as of Aug 15th)
- CDC Nationwide Commercial Laboratory Seroprevalence Survey estimated 3.9% [1.7% – 7.1%] seroprevalence as of Aug 15th (still no updates on these data from CDC)

These findings are equivalent to an ascertainment ratio of ~3x, with bounds of (1x to 7x)

- Thus for 3x there are 3 total infections in the population for every confirmed case
- Uncertainty design has been shifted to these bounds (previously higher ascensions as was consistent earlier in the pandemic were being used)



<https://covid.cdc.gov/covid-data-tracker/#national-lab>

Virginia Coronavirus Serology Project
Interim findings by region and statewide - July 22, 2020

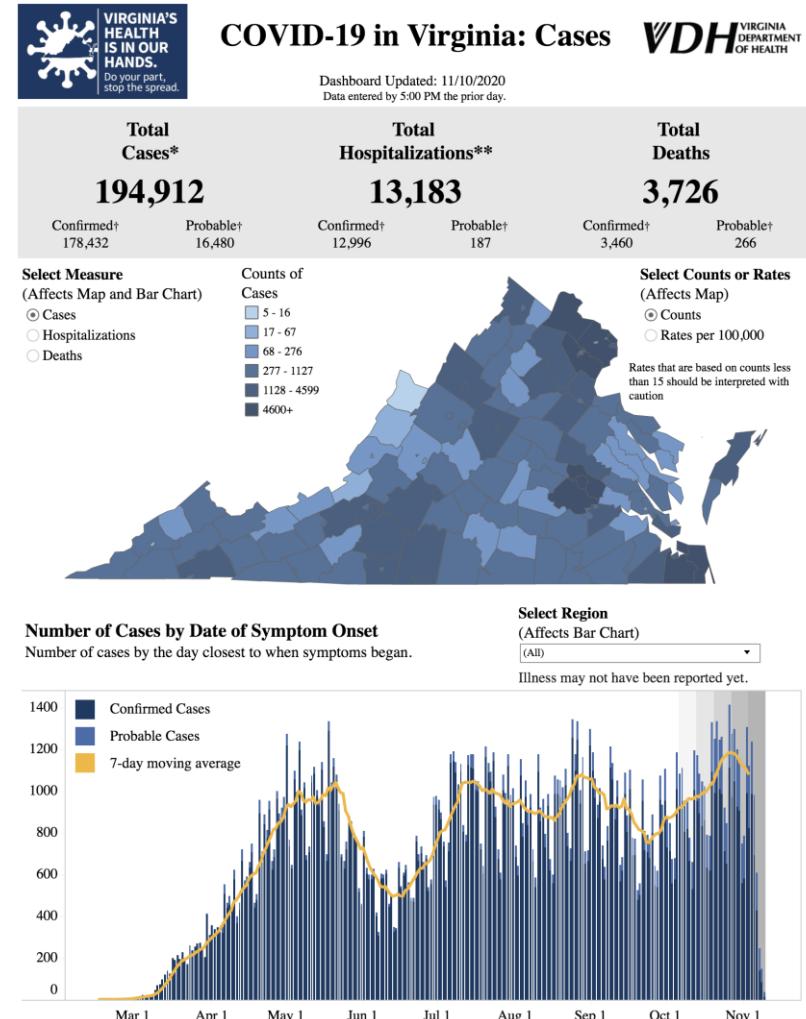
Region	Number of participants	Number antibody positive	Crude prevalence per 100 participants	Weighted prevalence*	
				per 100 population	(95% CI)
Central	400	8	2.0	3.0	(0.5, 5.5)
East	707	9	1.3	1.5	(-0.2, 3.2)
Northern	819	36	4.4	4.2	(2.5, 5.9)
Northwest	756	11	1.5	0.9	(0.2, 1.6)
Southwest	431	3	0.7	1.0	(-0.2, 2.1)
Virginia	3,113	67	2.2	2.4	(1.6, 3.1)

* Weighted prevalence is reweighted by region, age, sex, race, ethnicity, and insurance status to match census population.

<https://www.vdh.virginia.gov/content/uploads/sites/8/2020/08/VDH-Serology-Projects-Update-8-13-2020.pdf>

Calibration Approach

- **Data:**
 - County level case counts by date of onset (from VDH)
 - Confirmed cases for model fitting
- **Calibration:** fit model to observed data
 - Tune transmissibility across ranges of:
 - Duration of incubation (5-9 days), infectiousness (3-7 days)
 - Undocumented case rate (1x to 7x) guided by seroprevalence studies
 - Detection delay: exposure to confirmation (4-12 days)
 - Approach captures uncertainty, but allows model to precisely track the full trajectory of the outbreak
- **Project:** future cases and outcomes using the most recent parameters with constraints learned from the history of the fit parameters
 - Mean trend from last 7 days used, adjusted by variances in the previous 3 weeks
 - 1 week interpolation to smooth transitions in rapidly changing trajectories
 - Particles with high error or variance filtered out



Scenarios – Seasonal Effects

- Societal changes in the past month have led to an increase in transmission rates, these could continue to drive transmission
 - Seasonal impact of weather patterns
 - More interactions at places of learning
 - Travel related to holidays and traditional large family gatherings
 - Fatigue with infection control practices
- Population's behaviors determine the level of control of transmission we can achieve
- Three scenarios capture possible trajectories starting Nov 26th, 2020
 - Adaptive: No change from base projection
 - Adaptive-MoreControl: 15% decrease in transmission starting Nov 26th, 2020
 - Adaptive-LessControl: 15% increase in transmission starting Nov 26th, 2020



Model Results



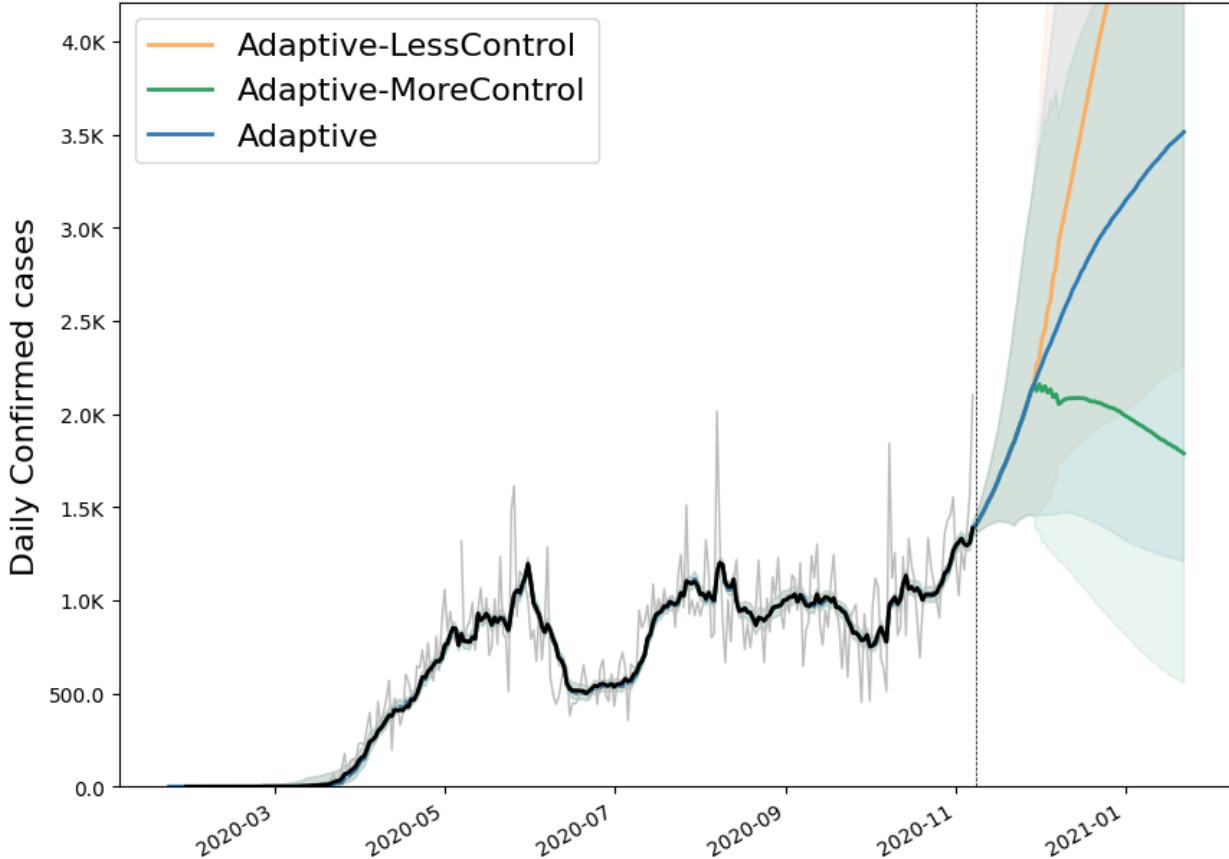
UNIVERSITY OF VIRGINIA

BIOCOMPLEXITY INSTITUTE

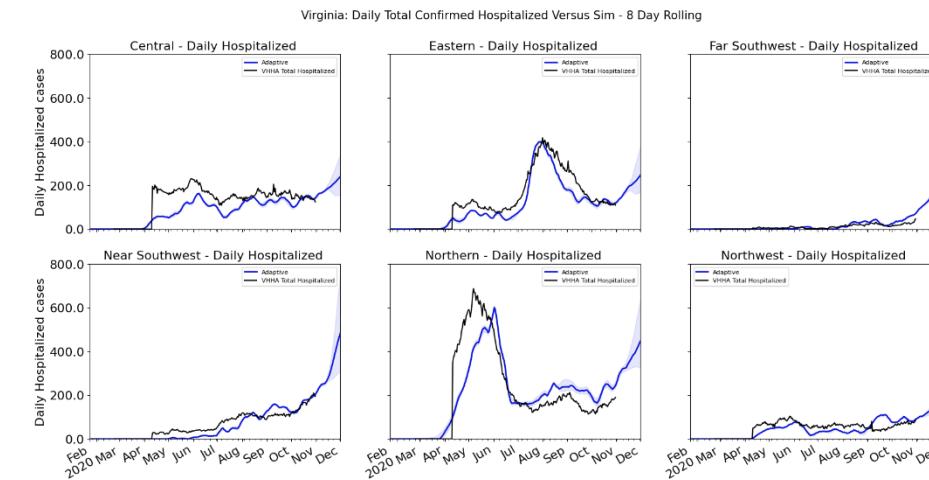
Outcome Projections

Confirmed cases

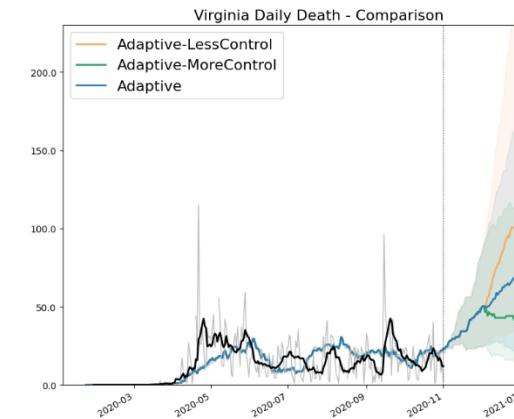
Virginia Daily Confirmed - Comparison



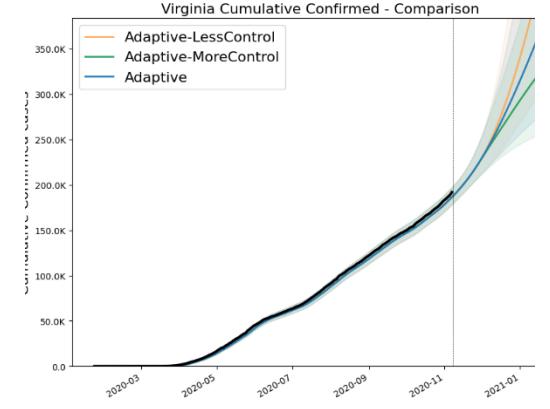
Estimated Hospital Occupancy



Daily Deaths



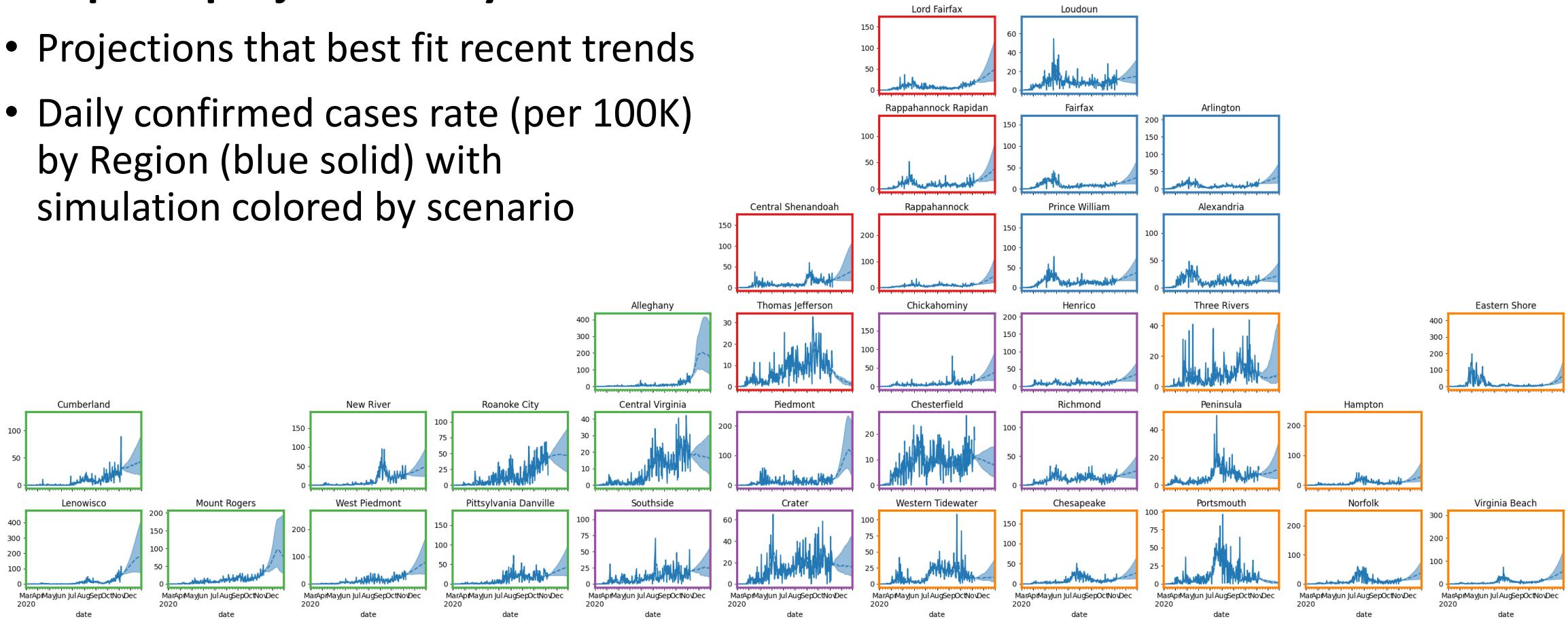
Cumulative Confirmed cases



District Level Projections: Adaptive

Adaptive projections by District

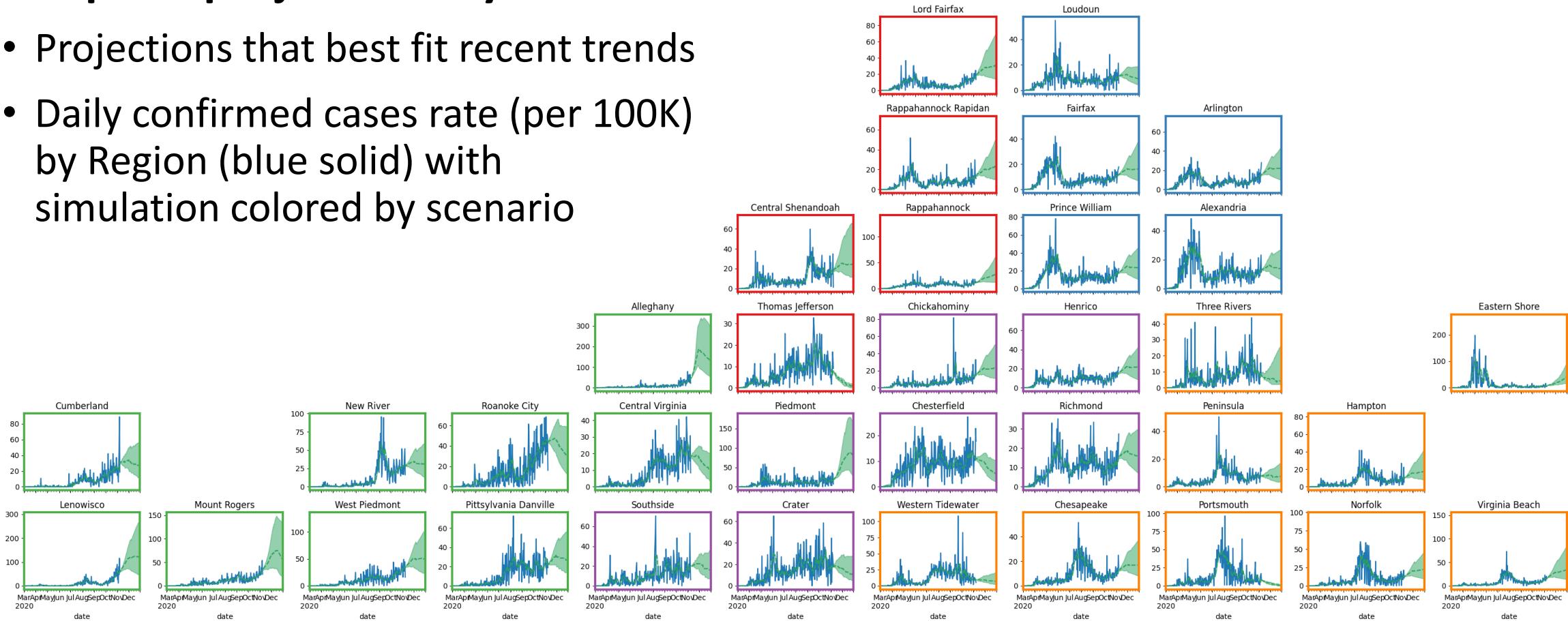
- Projections that best fit recent trends
- Daily confirmed cases rate (per 100K) by Region (blue solid) with simulation colored by scenario



District Level Projections: Adaptive-MoreControl

Adaptive projections by District

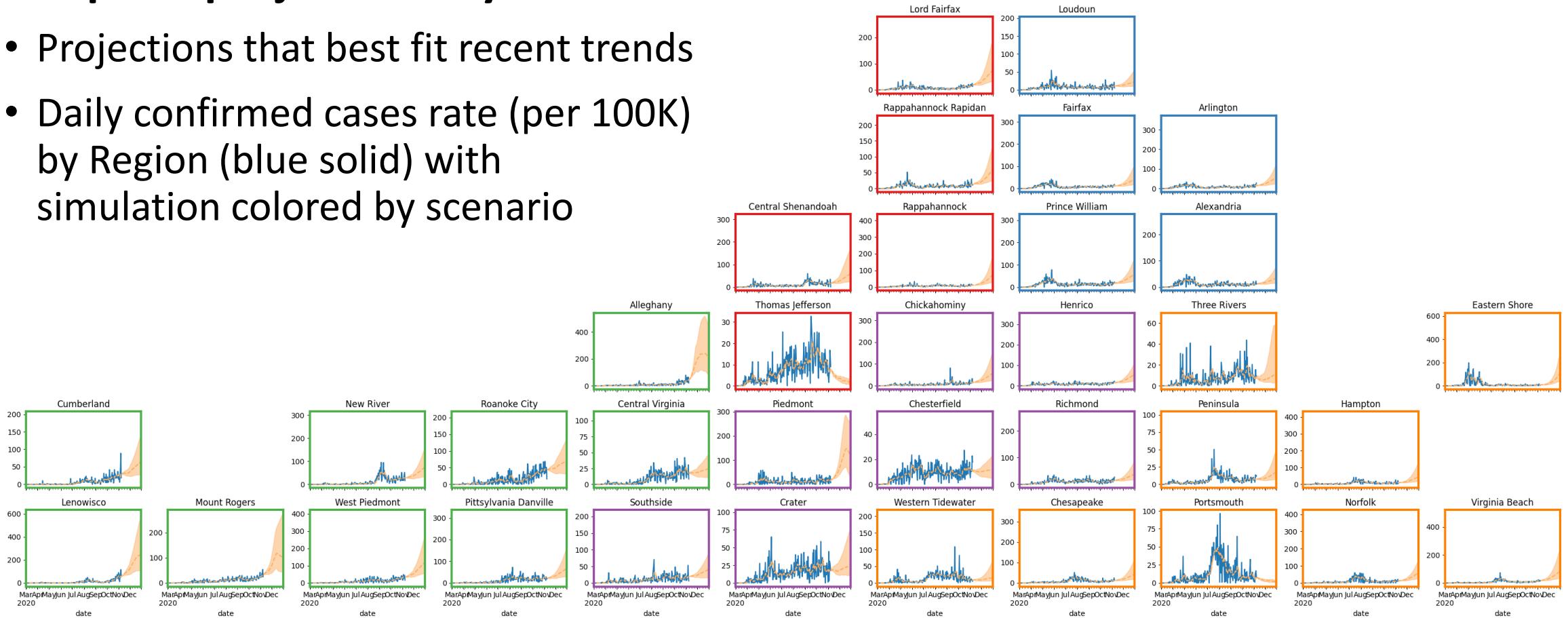
- Projections that best fit recent trends
- Daily confirmed cases rate (per 100K) by Region (blue solid) with simulation colored by scenario



District Level Projections: Adaptive-LessControl

Adaptive projections by District

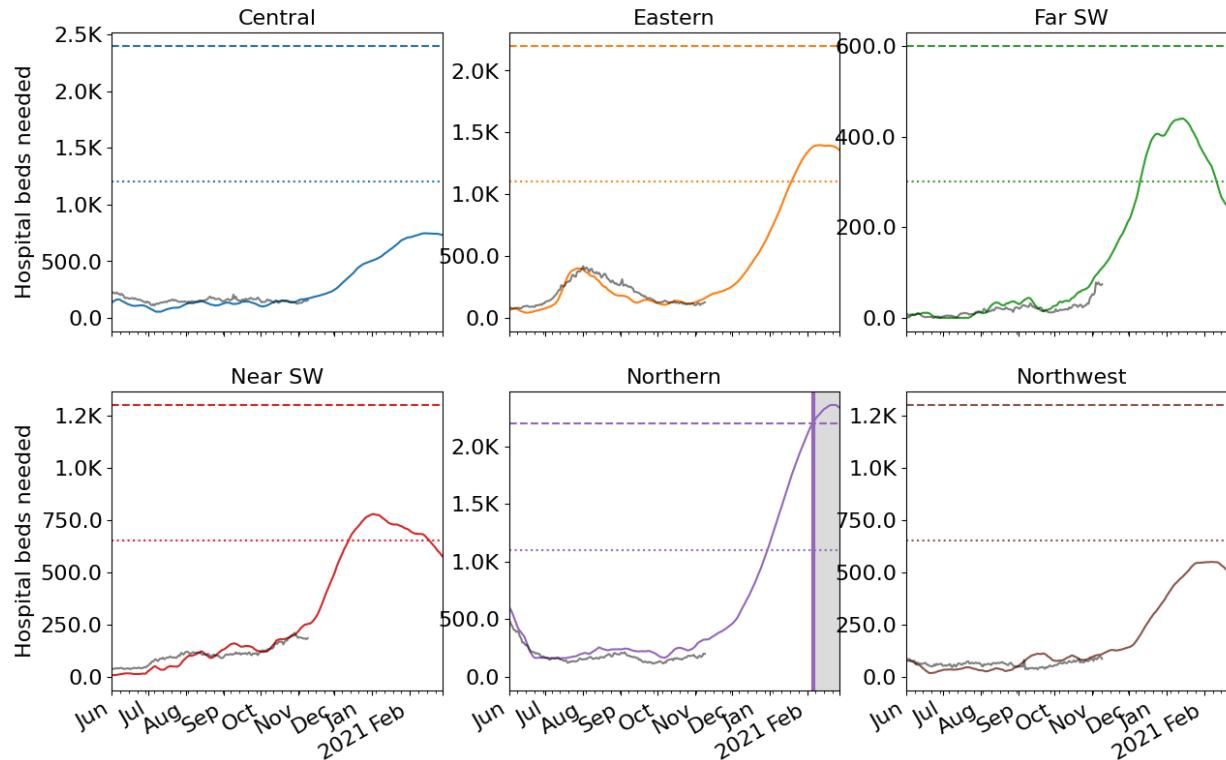
- Projections that best fit recent trends
- Daily confirmed cases rate (per 100K) by Region (blue solid) with simulation colored by scenario



Hospital Demand and Bed Capacity by Region

Capacities* by Region – Adaptive-LessControl

COVID-19 capacity ranges from 80% (dots) to 120% (dash) of total beds



Week Ending	Adaptive	Adaptive-LessControl
11/1/20	8,177	8,177
11/8/20	9,353	9,353
11/15/20	10,587	10,584
11/22/20	12,235	12,213
11/29/20	14,218	14,182
12/06/20	16,179	17,312
12/13/20	17,979	21,465
12/20/20	19,589	25,252
12/27/20	20,922	28,959
1/3/20	21,934	32,623
1/10/20	22,945	36,403
1/17/20	23,845	40,045

If Adaptive-LessControl scenario persists:

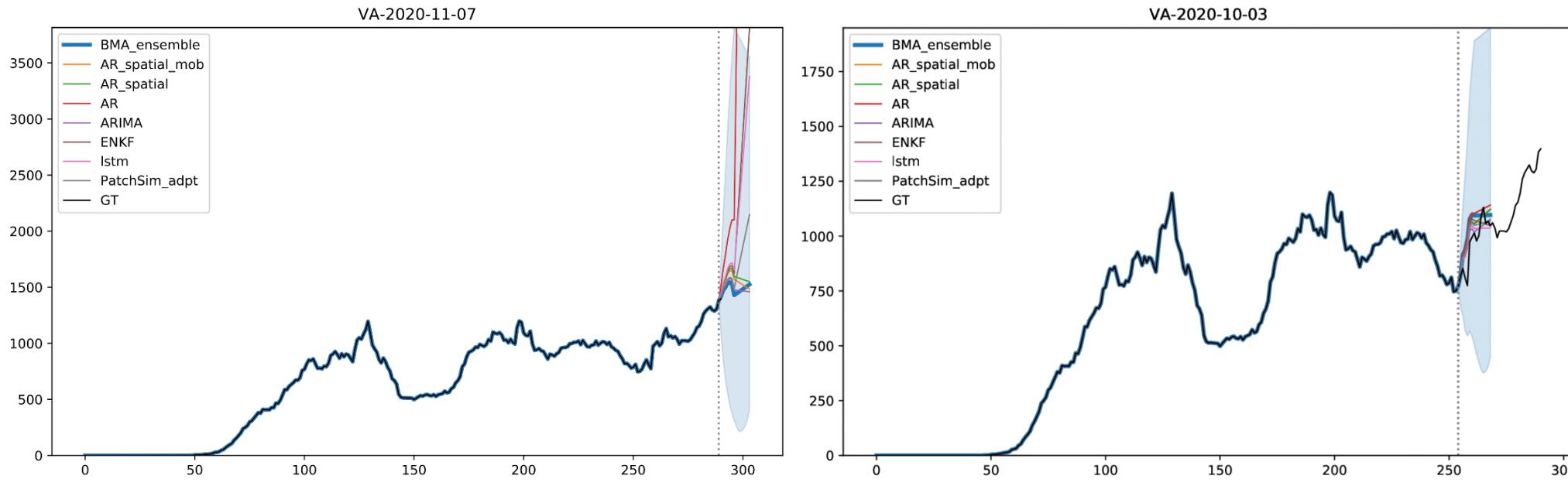
- Far and Near Southwest may begin to exceed initial capacity in early December, Eastern in January
- Northern could surpass initial capacity mid-Dec, surge capacity in early Feb

* Assumes average length of stay of 8 days

Additional Projection Methods Applied

An ensemble methodology that combines the Adaptive Fitting and other machine learning and statistical models has been developed to facilitate use of other data (weather, mobility, etc.)

- **Models:** Adaptive Fitting, ARIMA, LSTM, AR, spatially driven AR, Kalman Filters (ENKF)
- Ensemble will be folded into these projections after further training and evaluation



Key Takeaways

Projecting future cases precisely is impossible and unnecessary.

Even without perfect projections, we can confidently draw conclusions:

- **Virginia continues steady growth recording highest case rates of epidemic**
- VA mean weekly incidence (16.8/100K) is up again (from 14.8) though slower than nationally (46/100K from 34/100K).
- Projections are mostly up, showing potential for strain on health care system in some regions as early as December.
- Recent updates:
 - Horizon extended to March 1st
 - Preliminary results of ensemble of forecasting models included
 - Planning scenarios and case ascertainment rates remain as updated in previous weeks
- The situation is changing rapidly. Models will be updated regularly.

References

- Venkatramanan, S., et al. "Optimizing spatial allocation of seasonal influenza vaccine under temporal constraints." *PLoS computational biology* 15.9 (2019): e1007111.
- Arindam Fadikar, Dave Higdon, Jiangzhuo Chen, Bryan Lewis, Srinivasan Venkatramanan, and Madhav Marathe. Calibrating a stochastic, agent-based model using quantile-based emulation. *SIAM/ASA Journal on Uncertainty Quantification*, 6(4):1685–1706, 2018.
- Adiga, Aniruddha, Srinivasan Venkatramanan, Akhil Peddireddy, et al. "Evaluating the impact of international airline suspensions on COVID-19 direct importation risk." *medRxiv* (2020)
- NSSAC. PatchSim: Code for simulating the metapopulation SEIR model. <https://github.com/NSSAC/PatchSim> (Accessed on 04/10/2020).
- Virginia Department of Health. COVID-19 in Virginia. <http://www.vdh.virginia.gov/coronavirus/> (Accessed on 04/10/2020)
- Biocomplexity Institute. COVID-19 Surveillance Dashboard. <https://nssac.bii.virginia.edu/covid-19/dashboard/>
- Google. COVID-19 community mobility reports. <https://www.google.com/covid19/mobility/>
- Biocomplexity page for data and other resources related to COVID-19: <https://covid19.biocomplexity.virginia.edu/>



Questions?

Points of Contact

Bryan Lewis

brylew@virginia.edu

Srini Venkatramanan

srini@virginia.edu

Madhav Marathe

marathe@virginia.edu

Chris Barrett

ChrisBarrett@virginia.edu

Biocomplexity COVID-19 Response Team

Aniruddha Adiga, Abhijin Adiga, Hannah Baek, Chris Barrett, Golda Barrow, Richard Beckman, Parantapa Bhattacharya, Andrei Bura, Jiangzhuo Chen, Patrick Corbett, Clark Cucinell, Allan Dickerman, Stephen Eubank, Arindam Fadikar, Joshua Goldstein, Stefan Hoops, Ben Hurt, Sallie Keller, Ron Kenyon, Brian Klahn, Gizem Korkmaz, Vicki Lancaster, Bryan Lewis, Dustin Machi, Chunhong Mao, Achla Marathe, Madhav Marathe, Fanchao Meng, Henning Mortveit, Mark Orr, Joseph Outten, Akhil Peddireddy, Przemyslaw Porebski, SS Ravi, Erin Raymond, Jose Bayoan Santiago Calderon, James Schlitt, Aaron Schroeder, Stephanie Shipp, Samarth Swarup, Alex Telionis, Srinivasan Venkatramanan, Anil Vullikanti, James Walke, Amanda Wilson, Dawen Xie



UNIVERSITY OF VIRGINIA

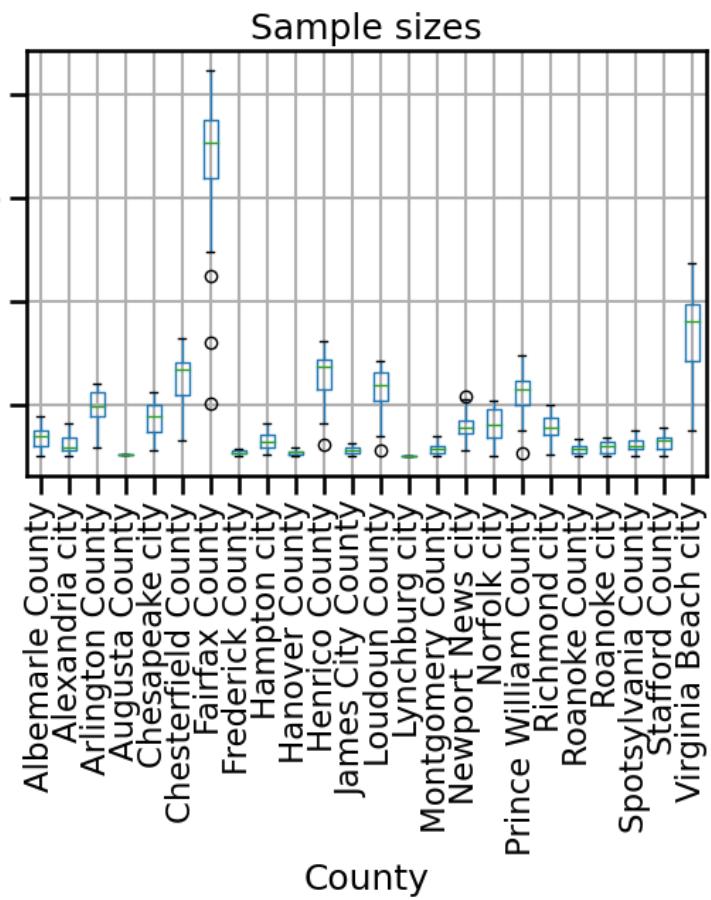
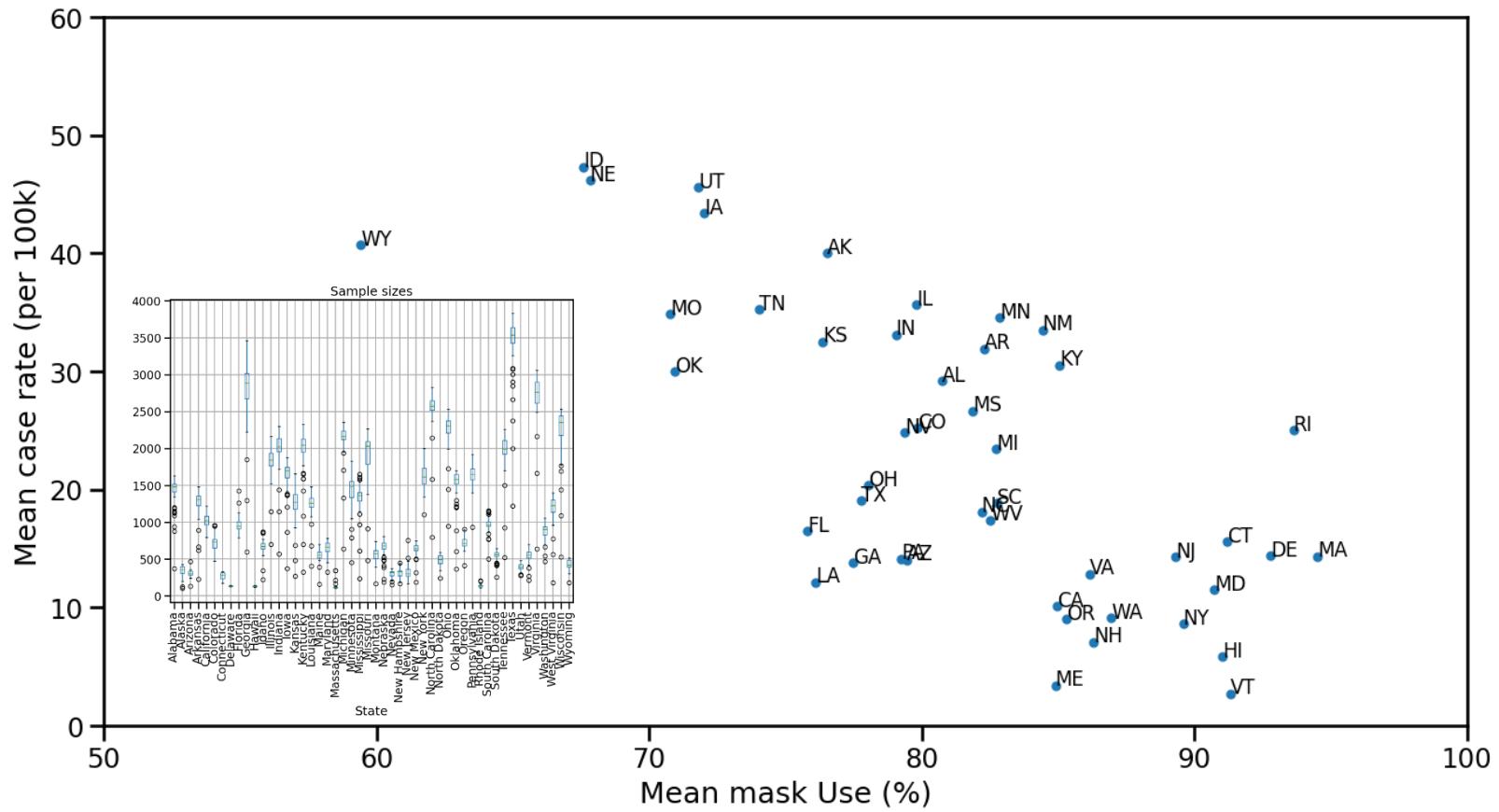
BIOCOMPLEXITY INSTITUTE

Supplemental Slides



BIOCOMPLEXITY INSTITUTE

Mask usage sample sizes



Test positivity across VA counties

- CMS weekly summary (used for guiding nursing homes testing protocol)
- Data: COVID-19 Electronic Lab Reporting (CELR); HHS Unified Testing Dataset;
- County level testing counts and test positivity rates for RT-PCR tests.
 - Green:** Test positivity <5.0% or with <20 tests in past 14 days
 - Yellow:** Test positivity 5.0%-10.0% or with <500 tests and <2000 tests/100k and >10% positivity over 14 days
 - Red:** >10.0% and not meeting the criteria for "Green" or "Yellow"

<https://data.cms.gov/stories/s/q5r5-gjyu>

County	Oct-14	Oct-21	Oct-28	Nov-04
Alleghany County	Green	Yellow	Red	Red
Bedford County	Red	Red	Red	Red
Botetourt County	Yellow	Yellow	Red	Red
Bristol City	Red	Red	Red	Red
Buckingham County	Yellow	Green	Yellow	Red
Campbell County	Red	Red	Red	Red
Carroll County	Yellow	Yellow	Red	Red
Charlotte County	Red	Red	Red	Red
Covington City	Yellow	Green	Red	Red
Craig County	Red	Red	Red	Red
Franklin City	Red	Red	Red	Red
Franklin County	Red	Red	Red	Red
Galax City	Yellow	Red	Red	Red
Giles County	Yellow	Yellow	Yellow	Red
Grayson County	Yellow	Yellow	Red	Red
Henry County	Red	Red	Red	Red
Lee County	Red	Red	Red	Red
Martinsville City	Red	Red	Red	Red
Prince George County	Red	Red	Red	Red
Prince William County	Yellow	Yellow	Red	Red
Pulaski County	Green	Yellow	Red	Red
Radford City	Yellow	Red	Red	Red
Roanoke City	Yellow	Yellow	Red	Red
Roanoke County	Red	Red	Red	Red
Rockingham County	Red	Yellow	Yellow	Red
Salem City	Yellow	Yellow	Red	Red
Scott County	Red	Red	Red	Red
Tazewell County	Yellow	Red	Red	Red
Washington County	Red	Red	Red	Red
Wise County	Yellow	Red	Red	Red

Red on Nov 04 (latest)

County	Oct-14	Oct-21	Oct-28	Nov-04
Amherst County	Red	Red	Red	Yellow
Bedford County	Red	Red	Red	Red
Bristol City	Red	Red	Red	Red
Campbell County	Red	Red	Red	Red
Charlotte County	Red	Red	Red	Red
Craig County	Red	Red	Red	Red
Emporia City	Red	Yellow	Red	Yellow
Franklin City	Red	Red	Red	Red
Franklin County	Red	Red	Red	Red
Greensville County	Red	Red	Green	Green
Henry County	Red	Red	Red	Red
Lee County	Red	Red	Red	Red
Manassas City	Red	Red	Yellow	Yellow
Martinsville City	Red	Red	Red	Red
Pittsylvania County	Red	Yellow	Red	Yellow
Prince George County	Red	Red	Red	Red
Roanoke County	Red	Red	Red	Red
Rockingham County	Red	Yellow	Yellow	Red
Scott County	Red	Red	Red	Red
Southampton County	Red	Red	Red	Yellow
Suffolk City	Red	Yellow	Yellow	Yellow
Washington County	Red	Red	Red	Red

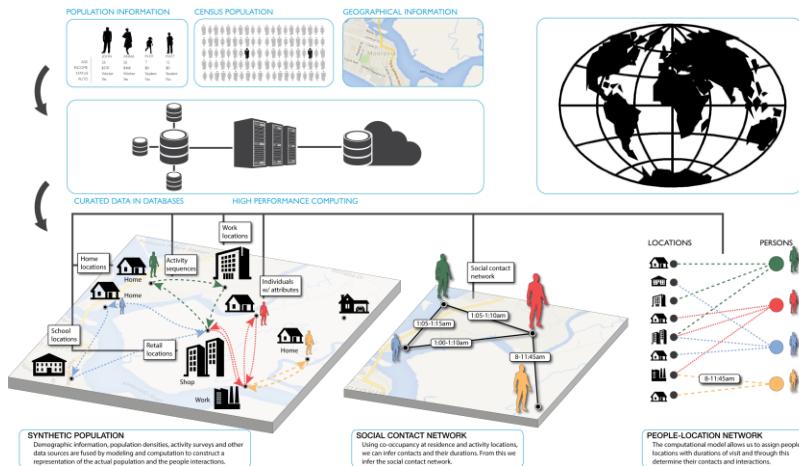
Red on Oct 14 (4-week back)



Agent-based Model (ABM)

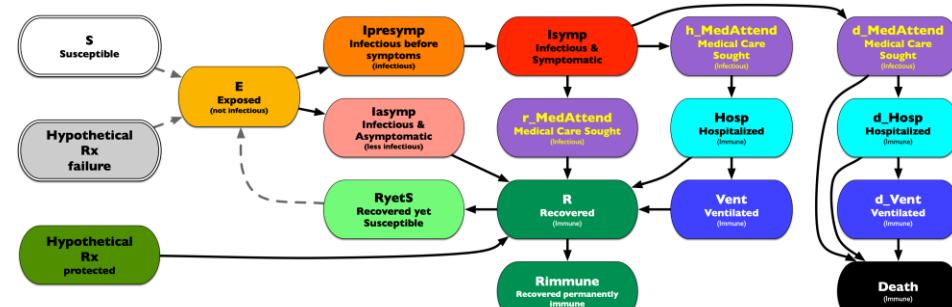
EpiHiper: Distributed network-based stochastic disease transmission simulations

- Assess the impact on transmission under different conditions
- Assess the impacts of contact tracing



Synthetic Population

- Census derived age and household structure
- Time-Use survey driven activities at appropriate locations



Detailed Disease Course of COVID-19

- Literature based probabilities of outcomes with appropriate delays
- Varying levels of infectiousness
- Hypothetical treatments for future developments

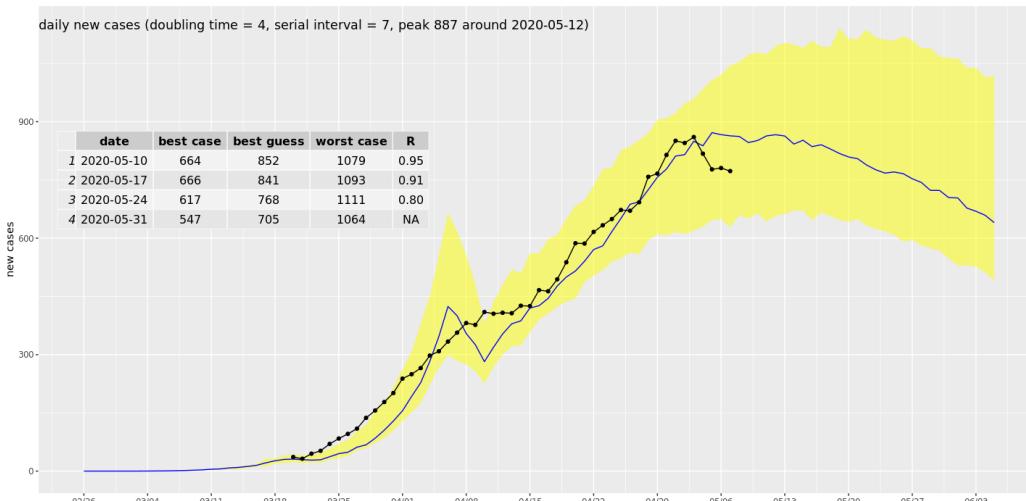


UNIVERSITY OF VIRGINIA

ABM Social Distancing Rebound Study Design

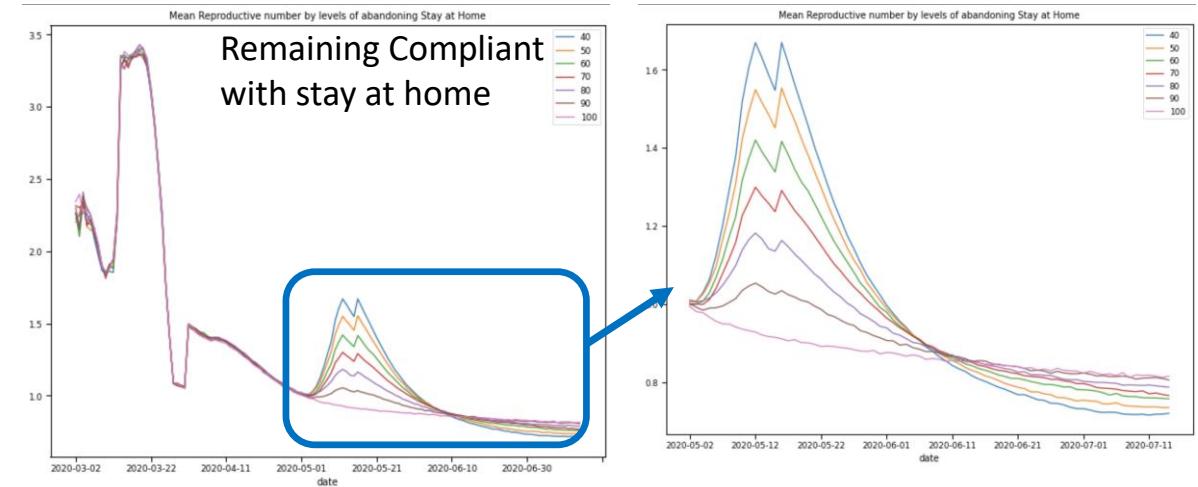
Study of "Stay Home" policy adherence

- Calibration to current state in epidemic
- Implement “release” of different proportions of people from “staying at home”



Calibration to Current State

- Adjust transmission and adherence to current policies to current observations
- For Virginia, with same seeding approach as PatchSim



Impacts on Reproductive number with release

- After release, spike in transmission driven by additional interactions at work, retail, and other
- At 25% release (70-80% remain compliant)
- Translates to 15% increase in transmission, which represents a 1/6th return to pre-pandemic levels

Medical Resource Demand Dashboard

<https://nssac.bii.virginia.edu/covid-19/vmrddash/>

