

Wastewater surveillance for COVID-19 forecasting

Promises and Pitfalls

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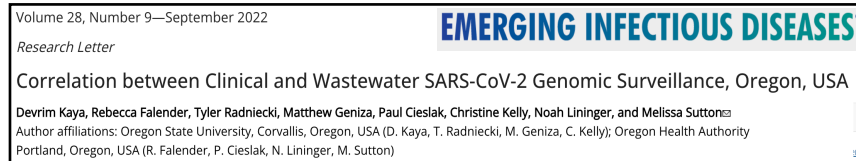
With feedback from VDH Office of Epidemiology

Motivation & Challenges

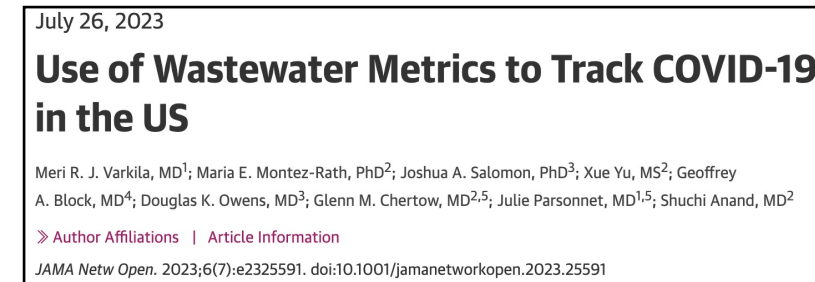
Detecting novel variants



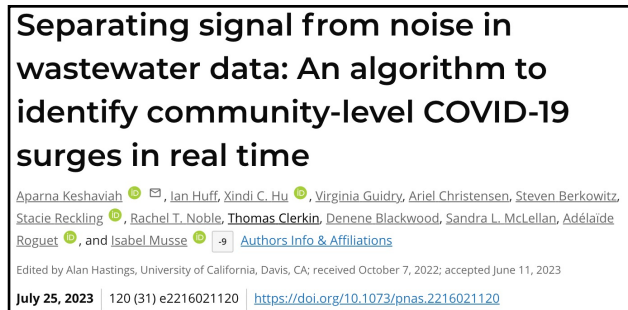
Tracking variant prevalence



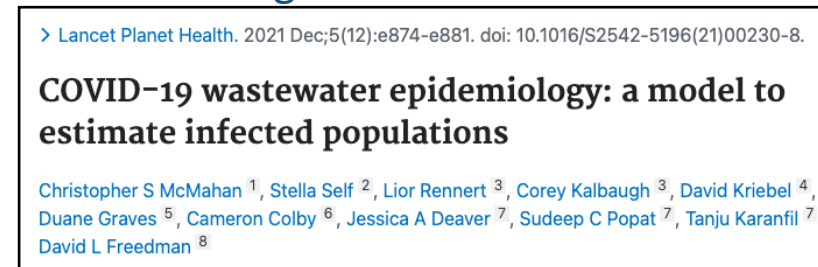
Association with high community incidence rates



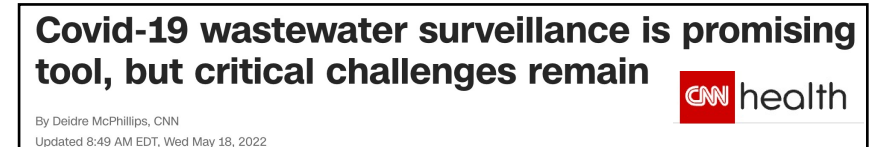
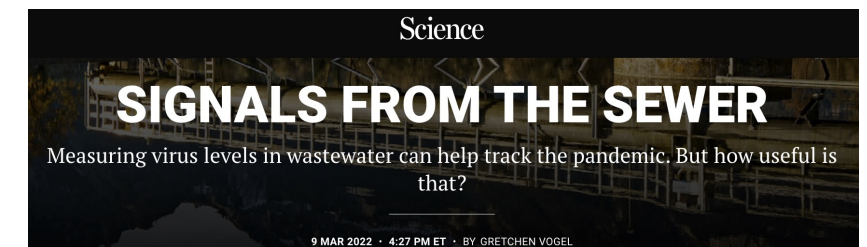
Early warning for community-level surges



Estimating true infection burden



<https://doi.org/10.17226/26767>

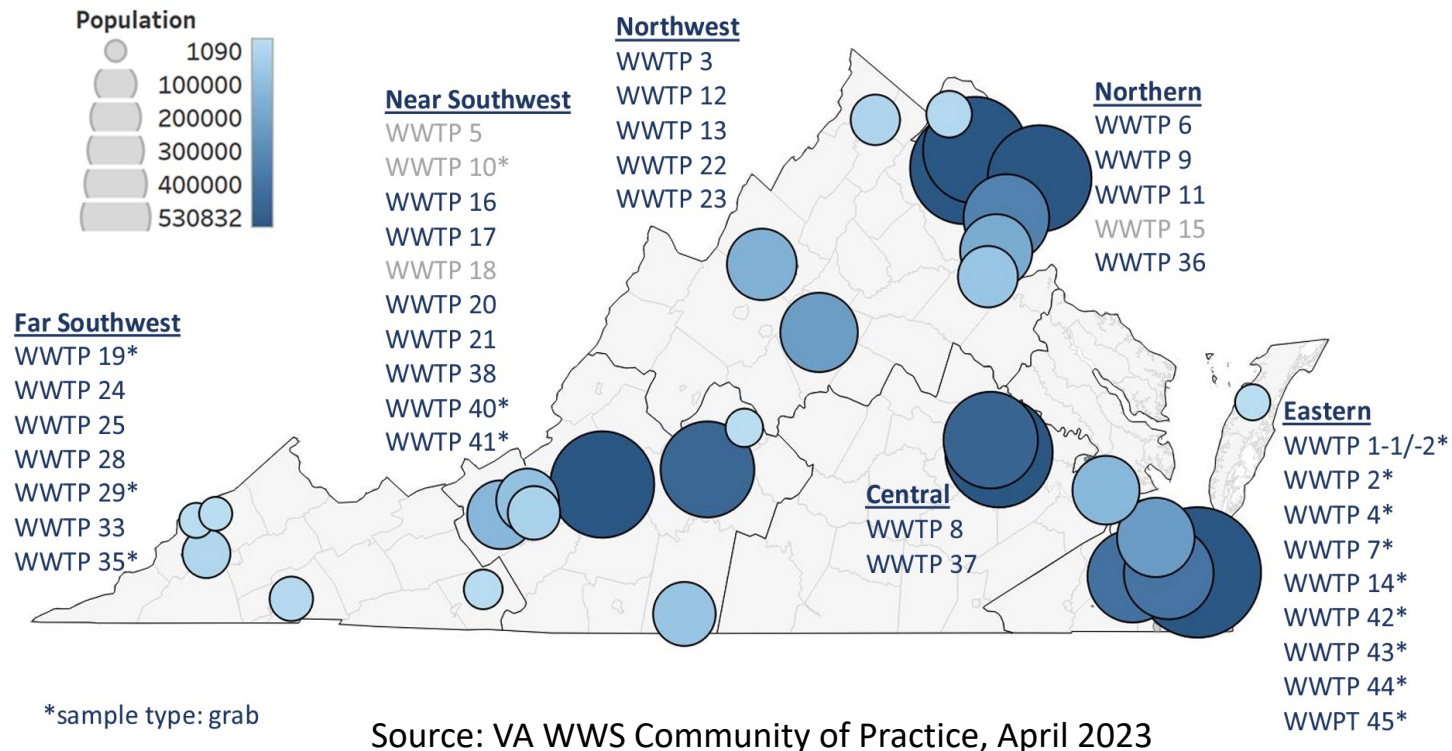


Key Questions

- How much *predictive power* does the wastewater surveillance time series (WWSTS) have for forecasting COVID-19 incidence?
- What are some of the *challenges* in working with WWSTS?
- What are some potential *approaches* to increase the utility of WWSTS?

Wastewater surveillance in Virginia

Sentinel Monitoring Facilities



Total 36 wastewater treatment plants (as of April 2023)

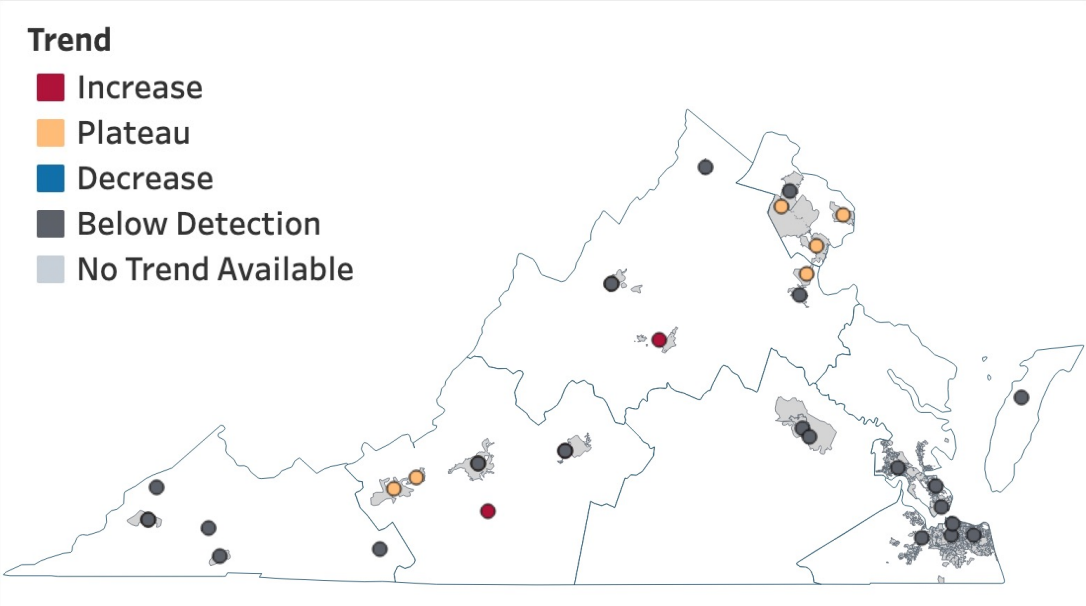
- 13 sampling twice weekly
- 23 sampling once weekly

Variant sequencing

- 20-23 (rotating) sites weekly

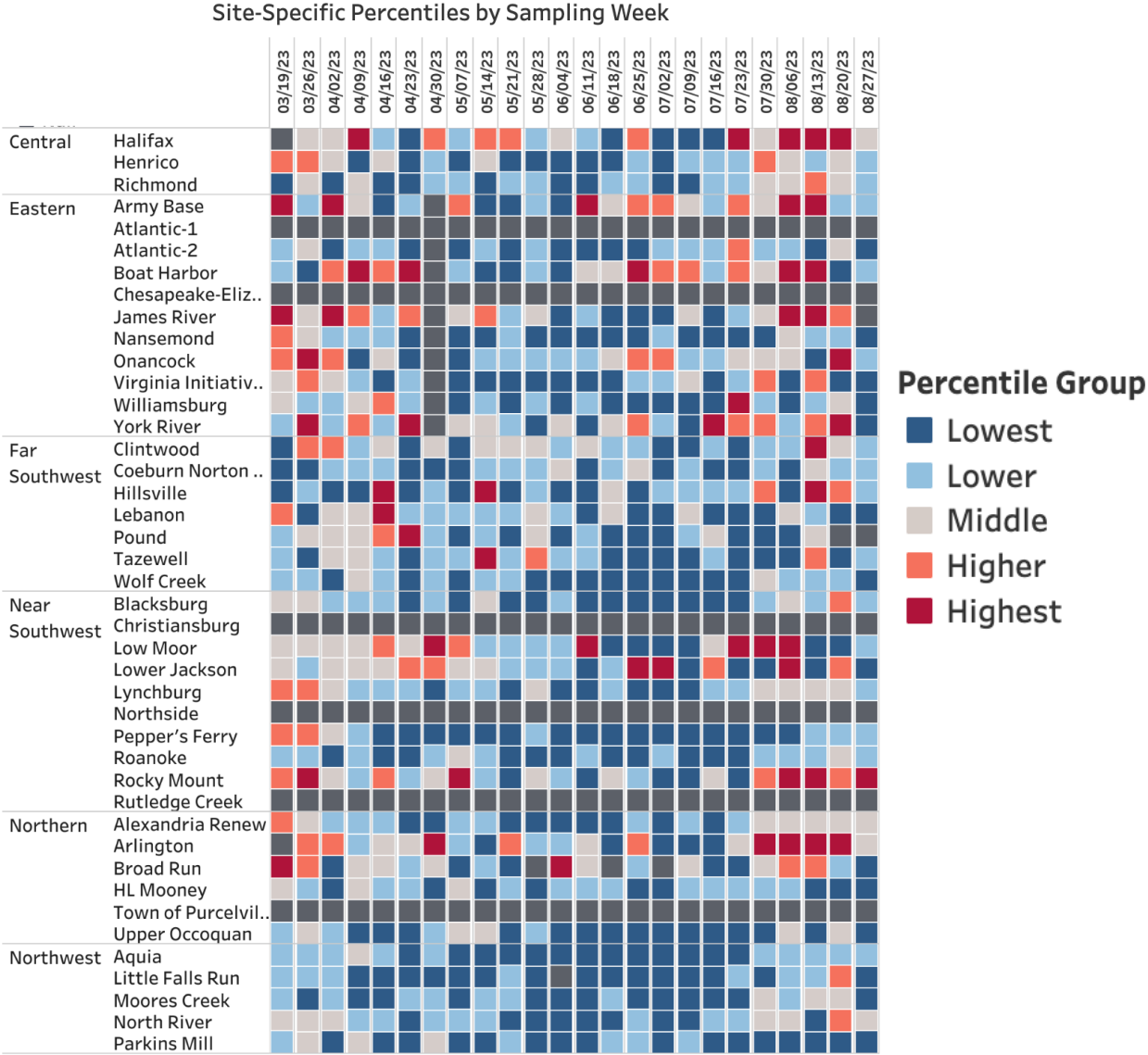
Public Communication

COVID-19 Wastewater Surveillance



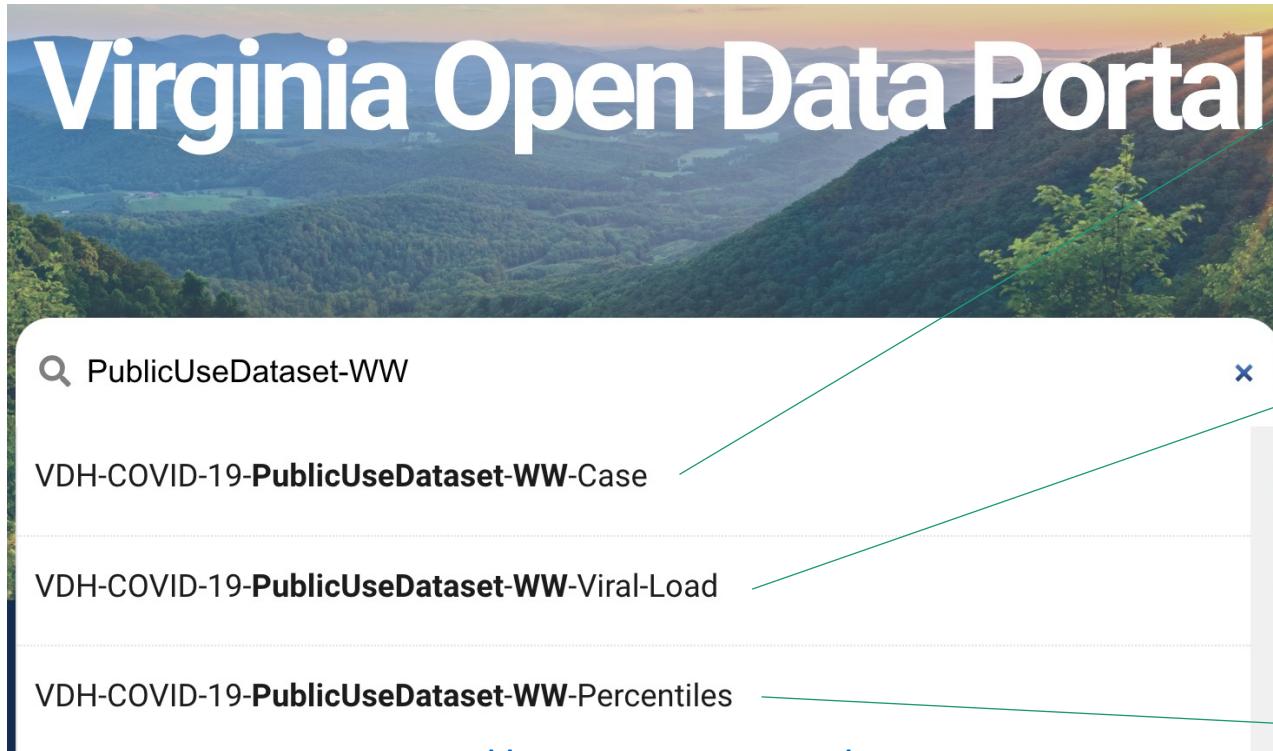
Trend Indicator

<https://www.vdh.virginia.gov/coronavirus/see-the-numbers/covid-19-data-insights/sars-cov-2-in-wastewater/>



Percentile Indicator

Public datasets



<https://data.virginia.gov/>

Cumulative cases by **sewershed (WWTP service area)** by event date

Weekly viral concentration by sewershed by sample collection week

Weekly percentile and percentile groups by sewershed by collection week

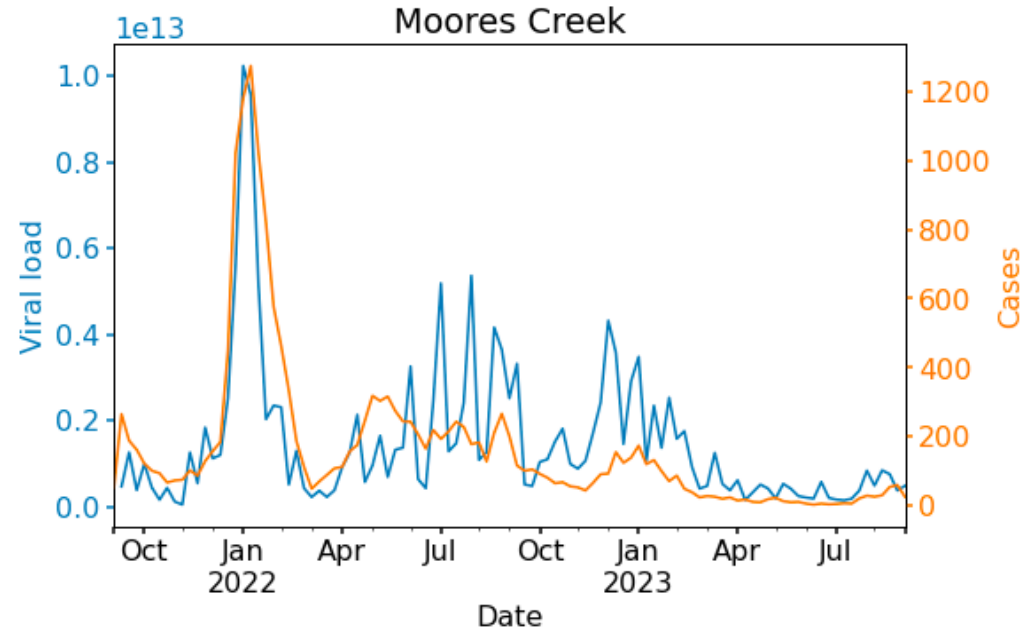
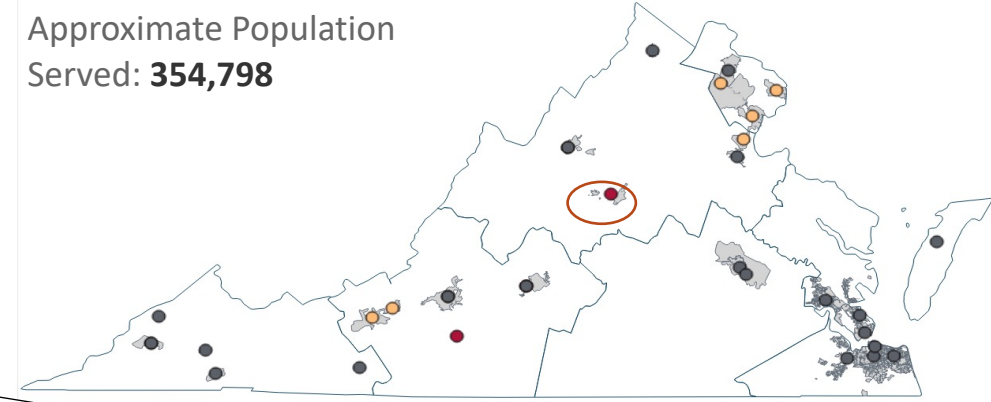
Additional data fields and interpretation

- Sampling methods:
 - Sample type: grab vs 24-hour composite
 - Sample matrix: raw wastewater vs post-grit removal
- Viral Measurements:
 - Limit of Detection for ddPCR analysis
 - Volumetric daily flow rate (million gallons/day)
 - Number of copies of PCR target per volume (copies/L)
 - Number of copies of PCR target per day (copies/day) ← Viral Load (VL)

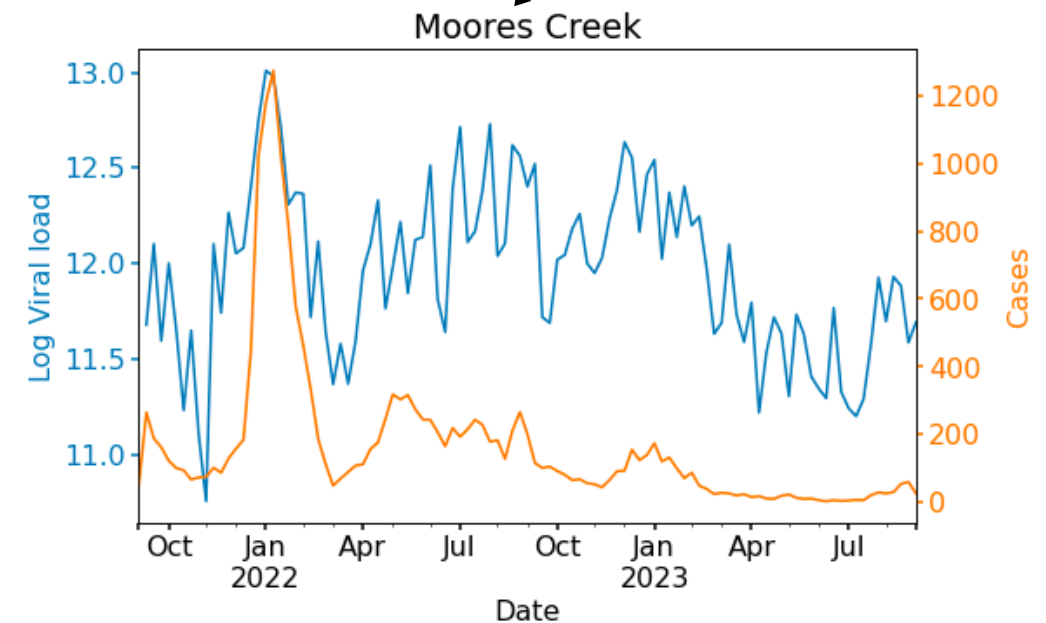
Visual comparison

Sewershed capturing Charlottesville-Albemarle counties

Sewershed: **Moore's Creek**
Approximate Population
Served: **354,798**



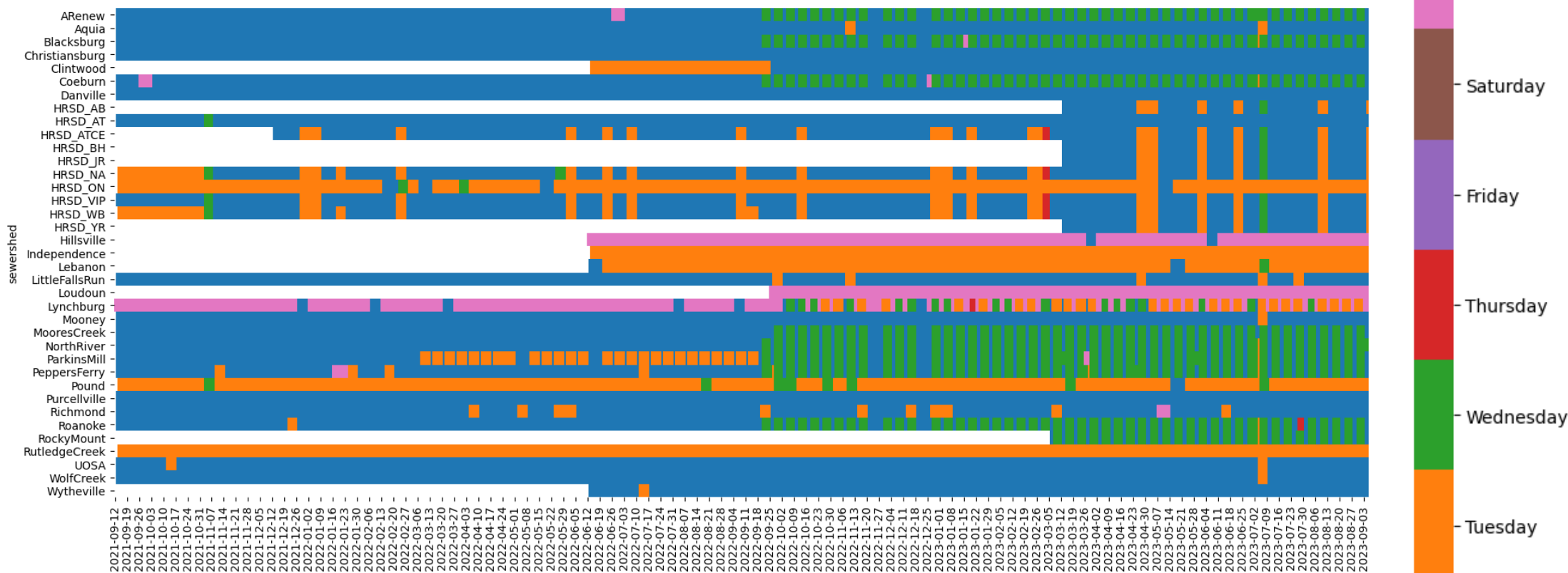
Viral load (copies/day)



After log transformation

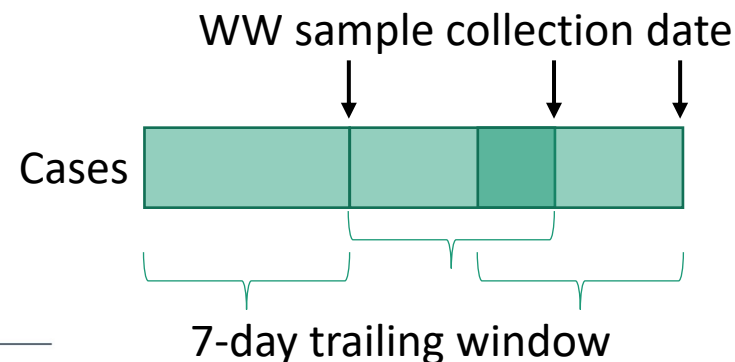
Temporal Alignment

Not all sites collect on the same day of week



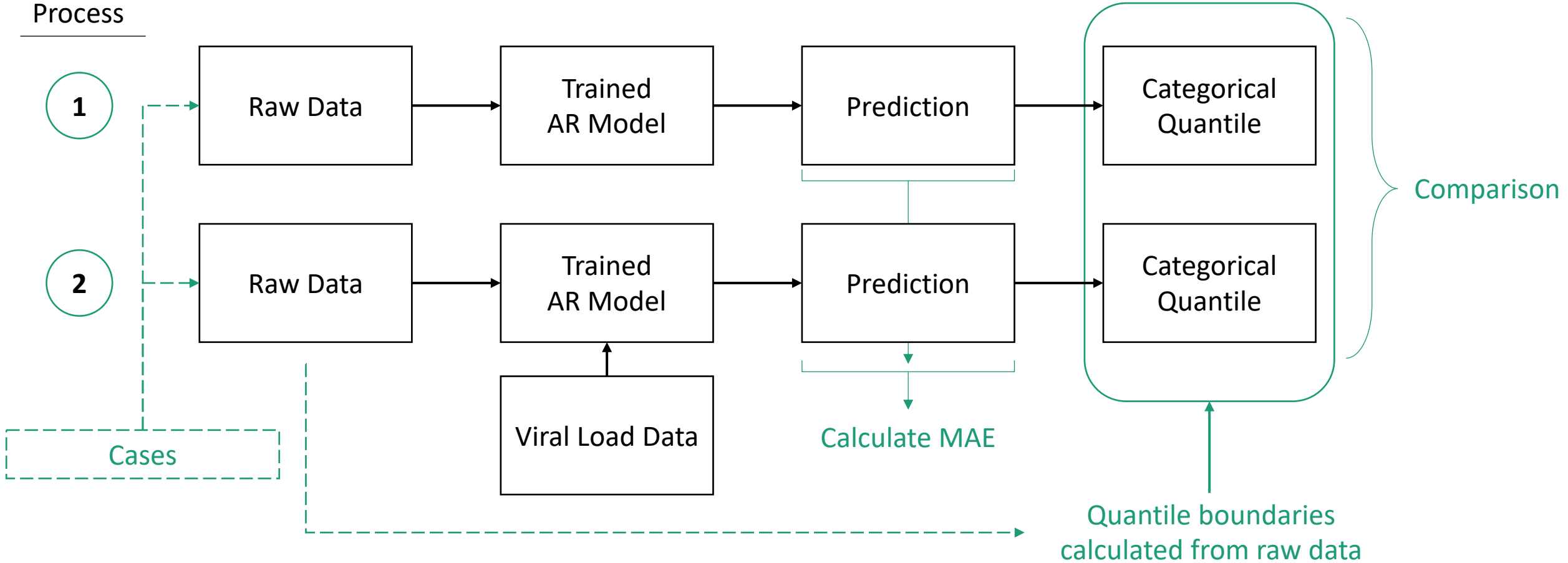
Aggregating daily cases (by event date)
in the 7 days before the collection date

$$c_w(t) = \sum_{i=0}^{p-1} c(t-i)$$

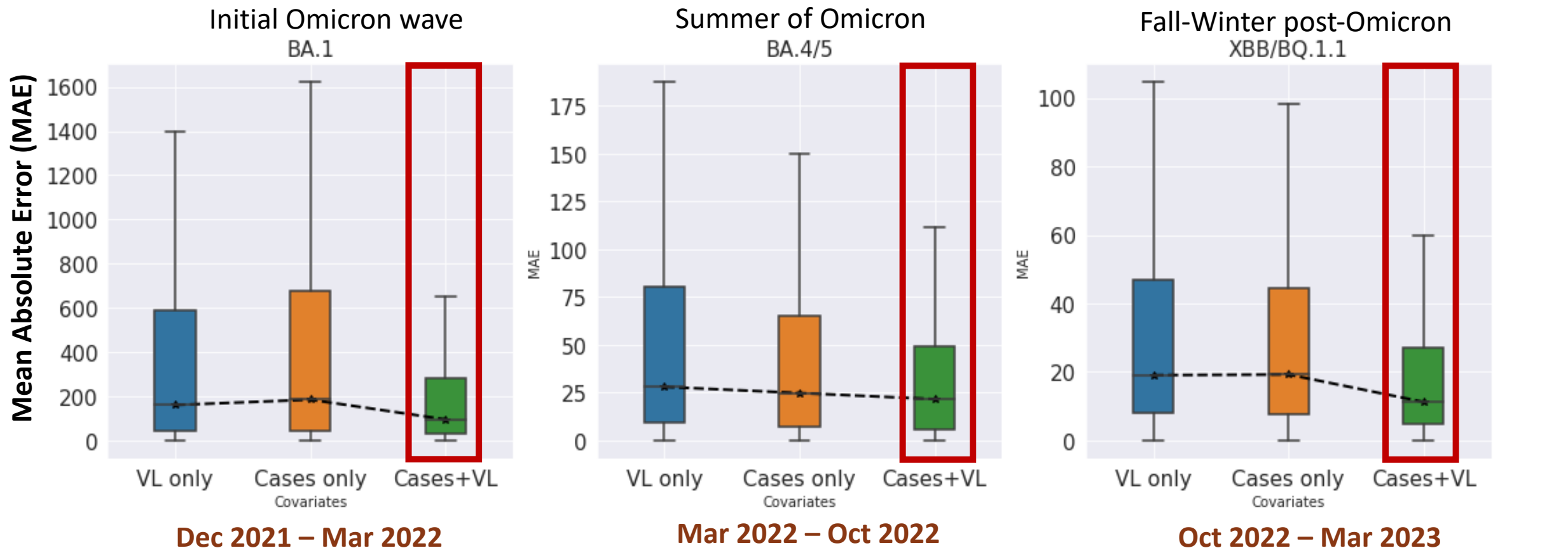


Analysis Framework

Process



Cases Prediction Performance (Across Waves)

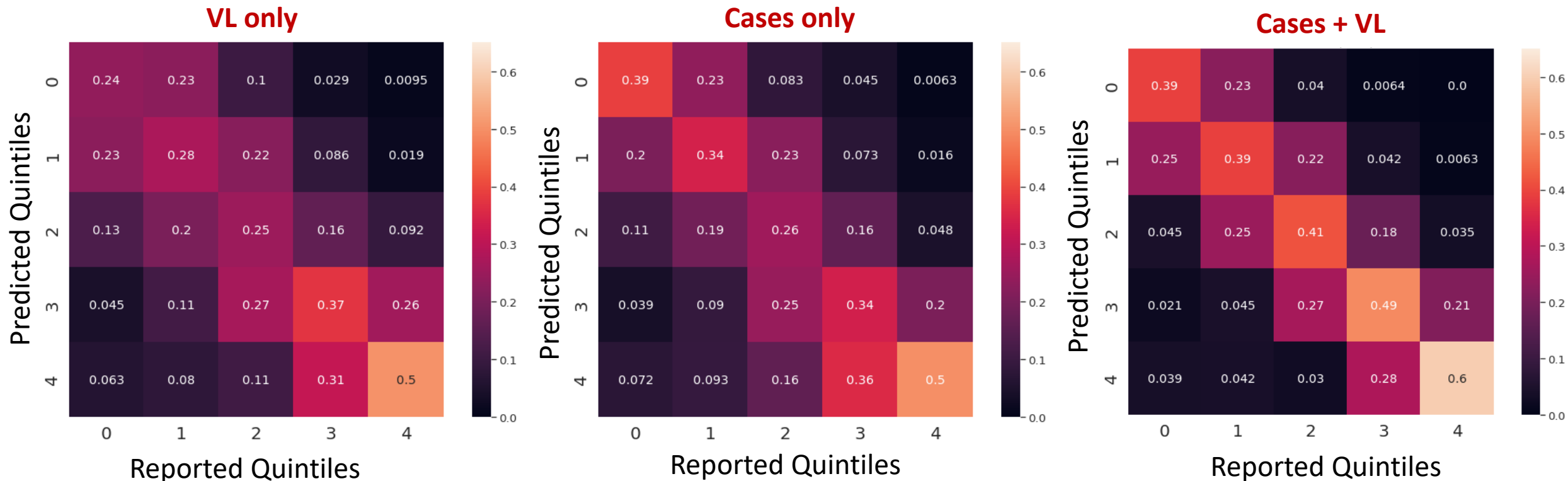


- Cases + VL model has lower median and IQR for MAE
- Performance improvement higher during winter waves

VL only: Model trained on Viral Loads only
Cases only: Model trained on Cases only
Cases+VL: Model trained on both Viral Load and Cases

Categorical comparison of predictions

- Convert case ground truth into quintiles
- Map each prediction into one of these bins (from ground truth)
- Better recall for highest quintiles; Cases + VL performs better

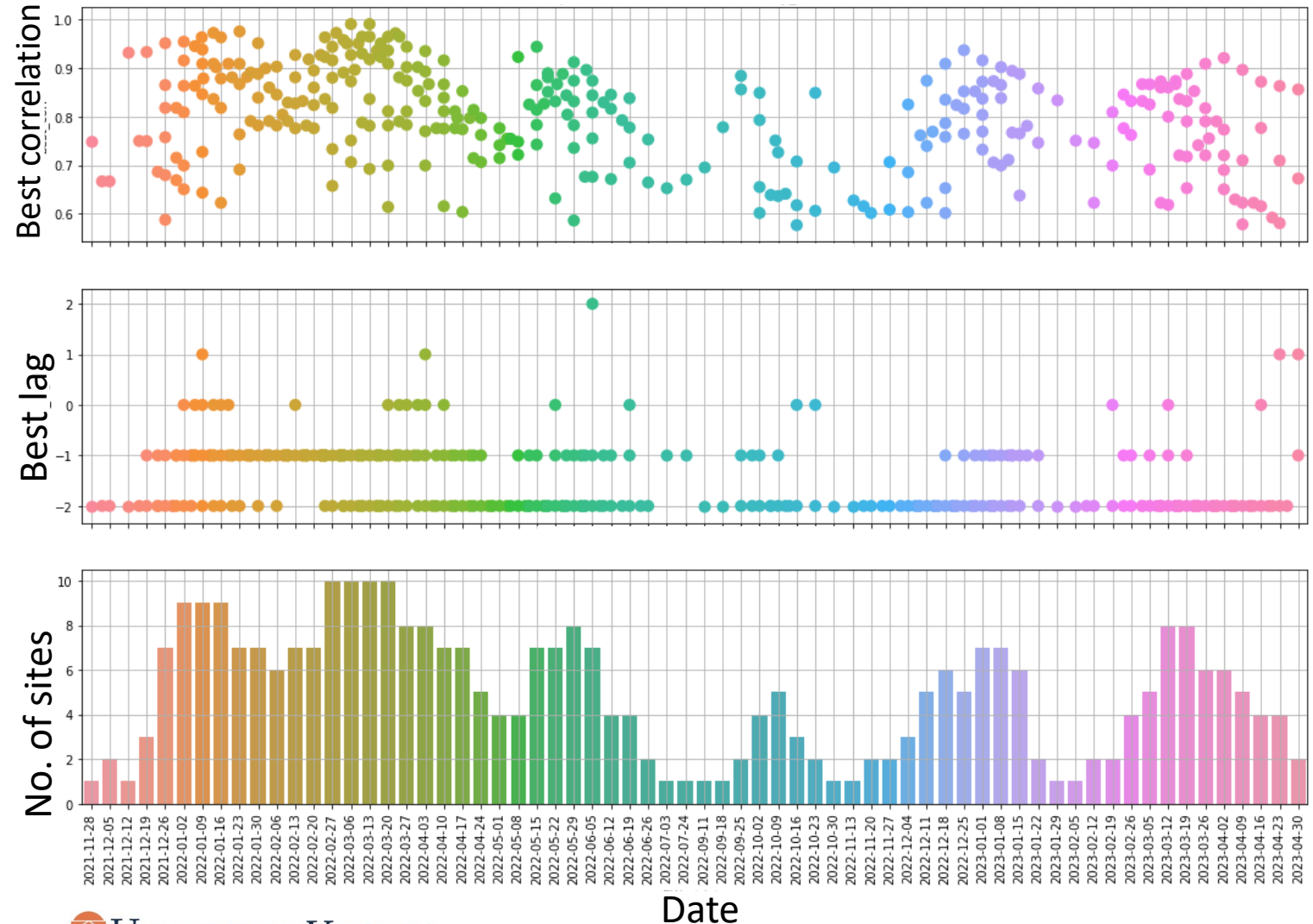


Correlation analyses at Sewershed level

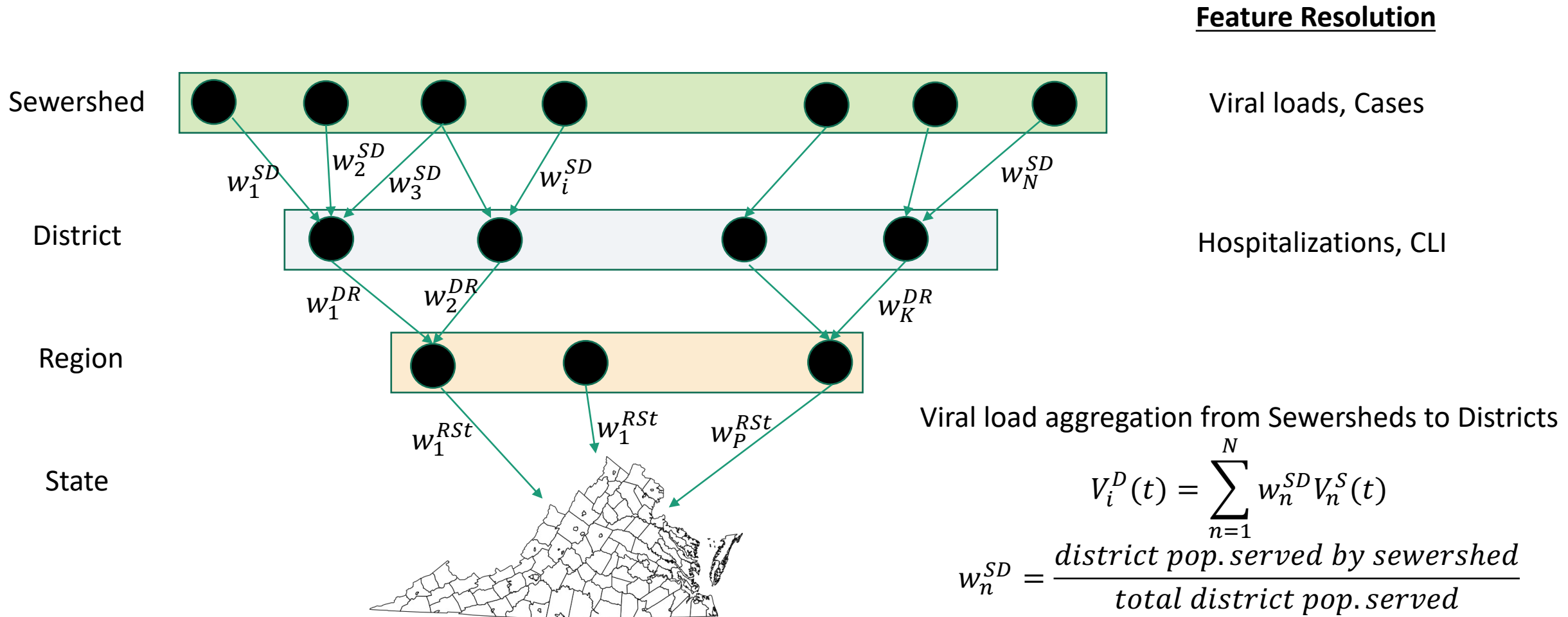
- A correlation-based lead-lag analysis with a rolling window of 12 weeks

$$\text{Corr}(\text{VL}(t+\text{lag}), \text{cases}(t))$$

- Variability in lag across time and sites, but mostly VL leads cases by 1-2 weeks
- The number of sites with statistically significant correlation increases during times of surge



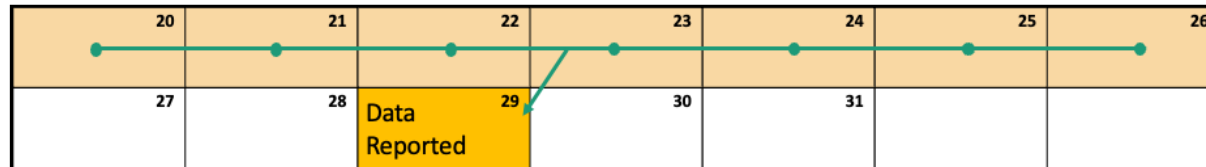
Need for Multi-Signal Approach



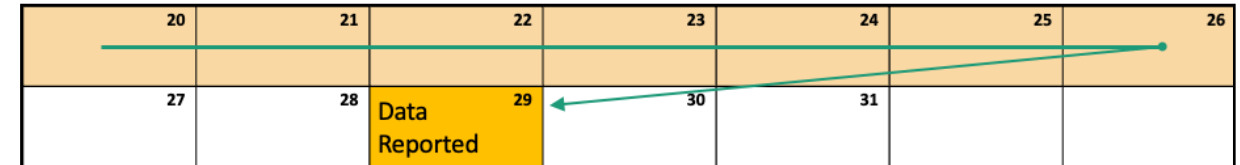
Aligning Reporting Cadences

“Latest” data available as of August 29th, 2023

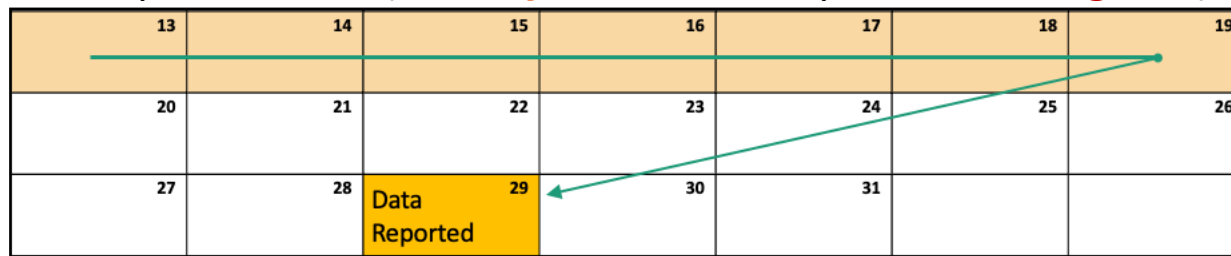
Cases (**Daily** data for the Epi week of **Aug 20th**)



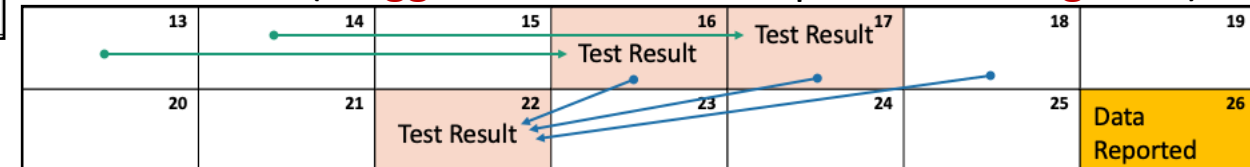
CLI (**Weekly** data for the Epi week of **Aug 20th**)



Hospitalizations (**Weekly** data for the Epi week of **Aug 13th**)



Wastewater (**Staggered** data for the Epi week of **Aug 13th**)



Key Takeaways

- Initial findings:

- Including viral load data in the model improves case forecasting accuracy
 - Performance improvement decays for subsequent Omicron waves
- High correlation and categorical alignment observed during peak activity

- Challenges:

- Need to account for variable sampling and signal noise
- Need to align with other signals across spatial resolutions
- Processing & reporting lags could impact real-time predictive power

- Ongoing work:

- Categorical forecasting of multiple signals with spatiotemporal alignment