

Seasonality of Respiratory Infections: From Laboratory to Model, Forecast, and Public Health Action

Wan Yang, PhD
Assistant Professor of Epidemiology

Pennsylvania State University CIDD Seminar
Sep 14, 2023

Outline

with a focus on influenza and COVID-19:

- ▶ **Epidemiology:** What is the seasonality of respiratory infections?
- ▶ **Laboratory studies:** How and why?
- ▶ **Modeling:** How to translate the mechanisms into models?
- ▶ **Forecasting:** Can seasonality models help forecast?
- ▶ **Public Health Action:** How all the above can help public health?

Acknowledgements

► Collaborators

○ Flu studies:

- ❖ Virginia Tech: Linsey Marr (phd work);
- ❖ Columbia University: Jeff Shaman, Max O'Donnell, Mat Cummings; Haokun Yuan, Sarah Kramer
- ❖ University of Hong Kong: Ben Cowling, Eric Lau
- ❖ Uganda Virus Research Institute: Barnabas Bakamutumaho, John Kayiwa, Nicholas Owor, Barbara Namagambo, Timothy Byaruhanga, Julius J. Lutwama

○ COVID studies:

- ❖ Columbia: Jeff Shaman, Sasi Kandula, Haokun Yuan
- ❖ NYC DOHMH: Sharon Greene, Anne Fine, Jaimie Shaff, and many others

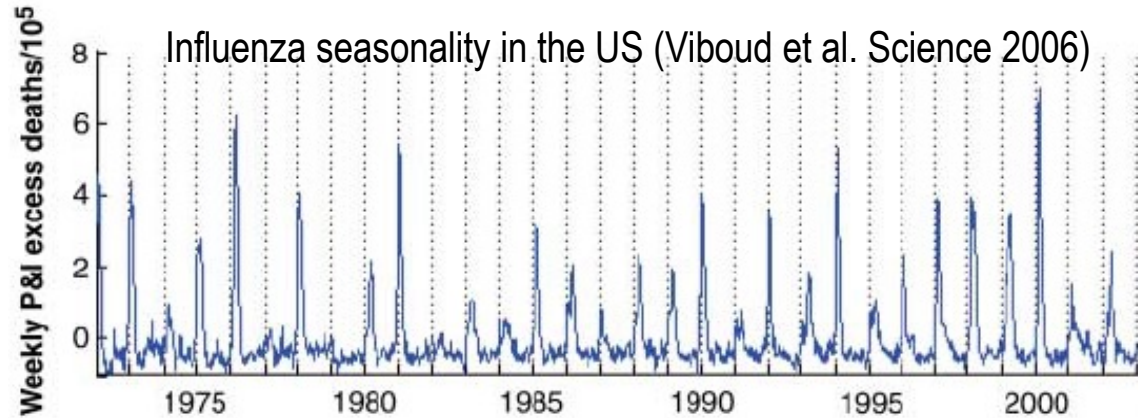
► Funding:

- NIH (AI145883; AI135926; ES009089)
- NSF (DMS-2027369)
- CDC/CSTE (NU38OT00297; 75D30122C14289)
- NYC DOHMH

RESPIRATORY INFECTION SEASONALITY: EPIDEMIOLOGY

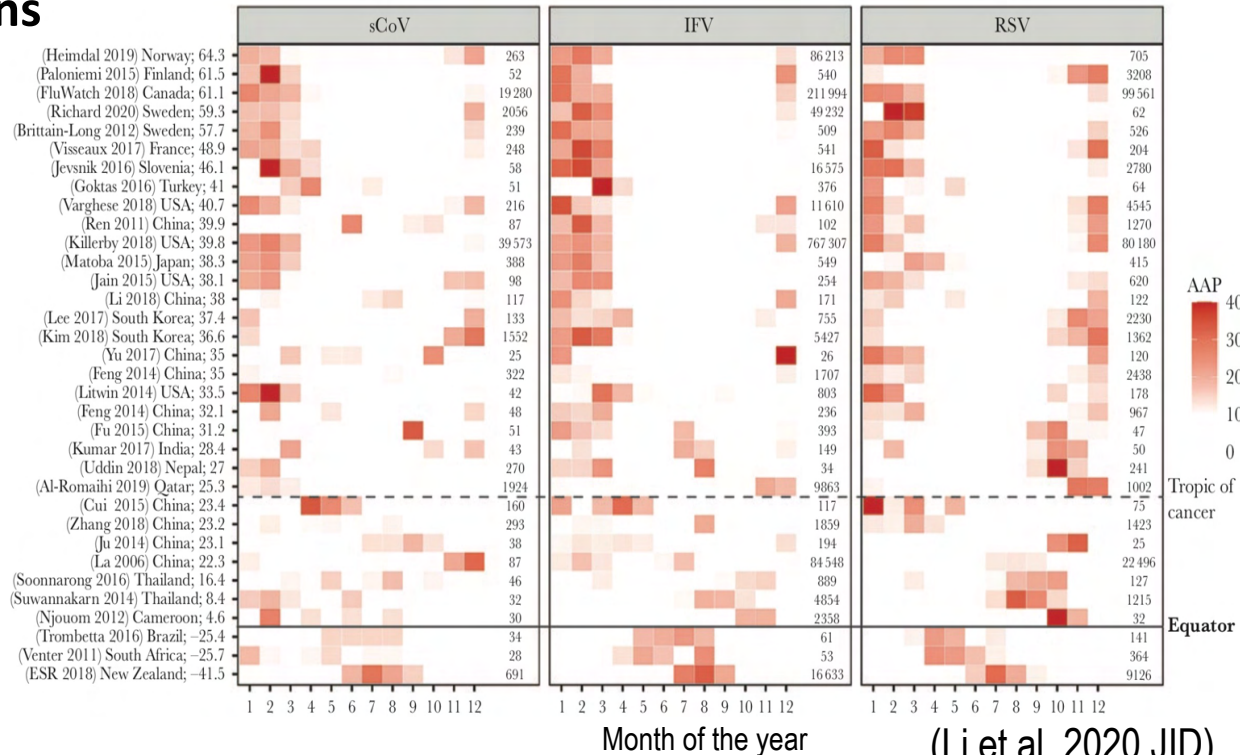
Seasonality of common respiratory infections

- ▶ Influenza tends to surge in the winter in temperate regions



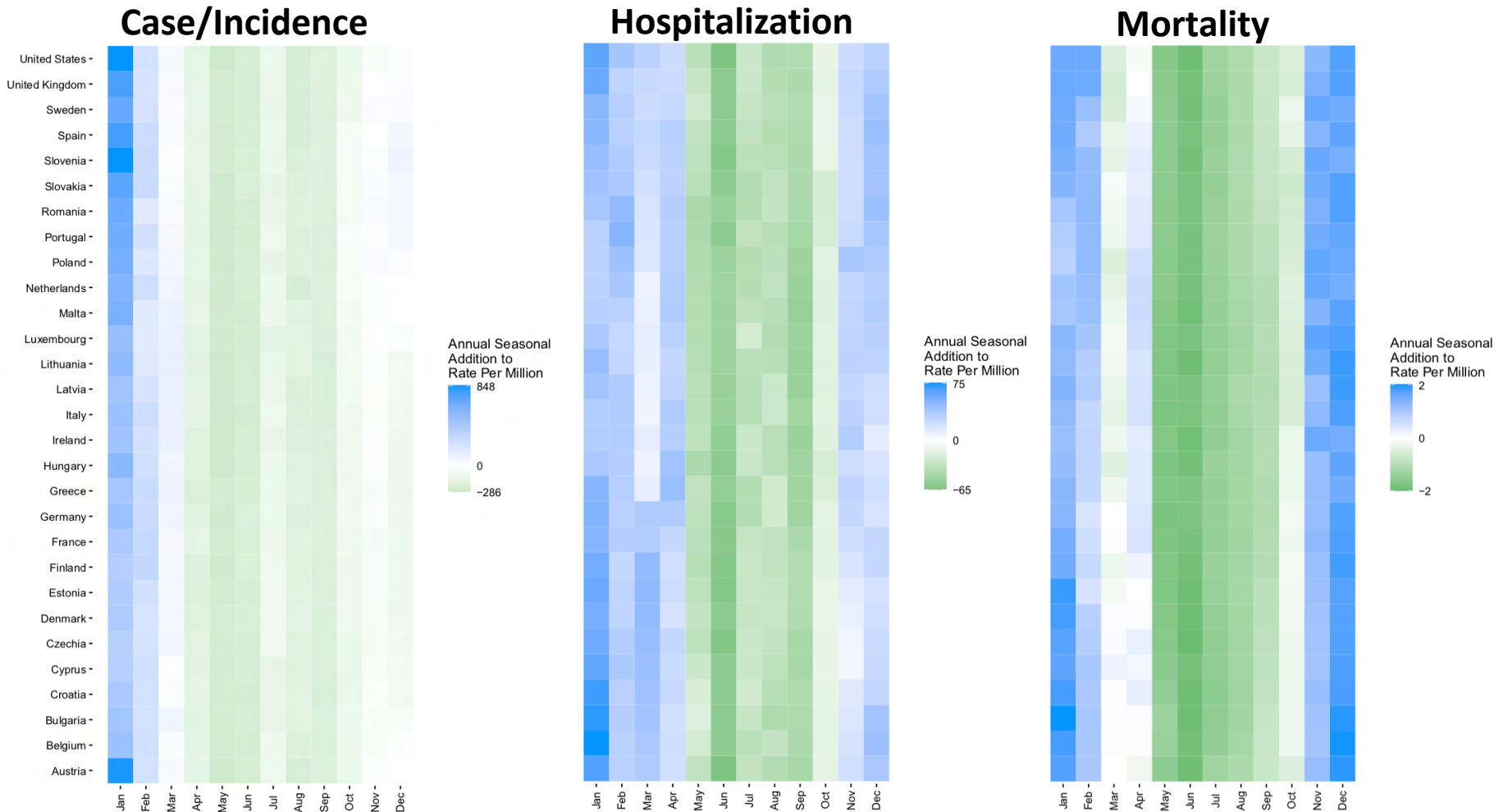
- ▶ Many respiratory infections seem to follow similar seasonal pattern:

- Temperate regions: cold winter-months
- Tropical/subtropical: two peaks, or year round (less studied)



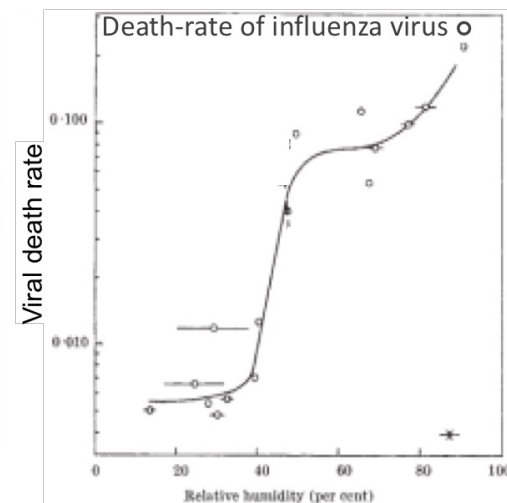
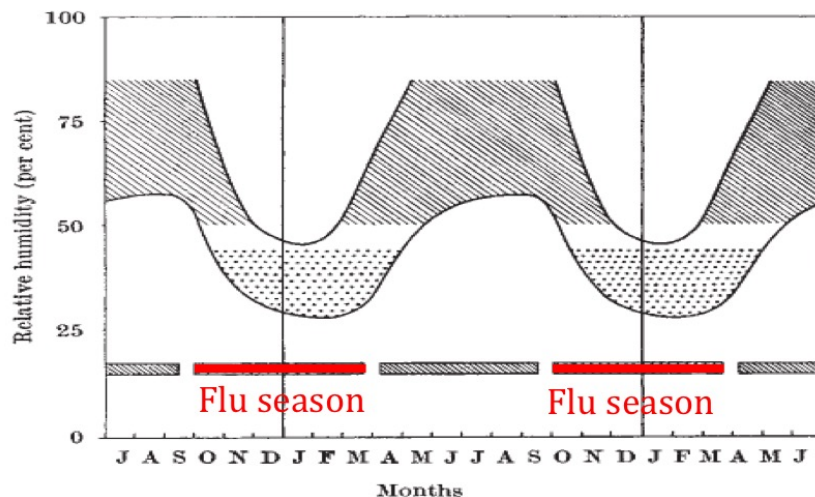
Seasonality of common respiratory infections

- ▶ **COVID-19/SARS-CoV-2: higher incidence, hospitalization, and mortality in the winter in temperate regions**



General mechanisms behind this seasonality

- ▶ **Three main mechanisms** (Moriyama et al. 2020 Annual Review of virology)
 - Host vulnerability
 - Host behavior
 - Viral stability (Environmental factors: temperature, humidity)
- ▶ **Flu seasonality v. humidity**
 - H: Low wintertime humidity drives flu epidemic (Hemmes et al. 1960)
 - ❖ Temperate regions: Flu epidemic coincides with low indoor humidity
 - ❖ Influenza virus survives better at lower humidity
 - *Supports the seasonality in temperate regions, but not the (sub)tropics*

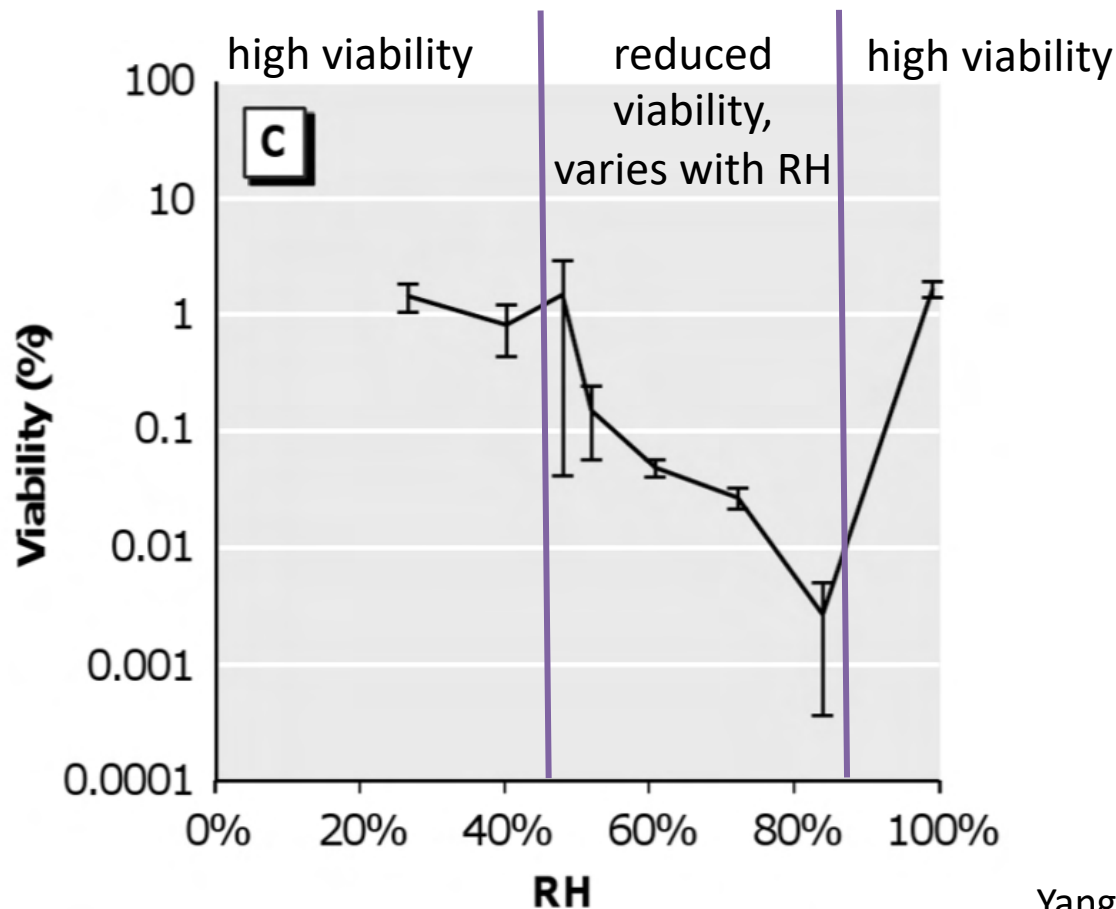


Hemmes et al. Nature 1960

(SELECT) LABORATORY STUDIES: ENVIRONMENTAL FACTORS, AND UNDERLYING MECHANISMS

Laboratory studies: Influenza

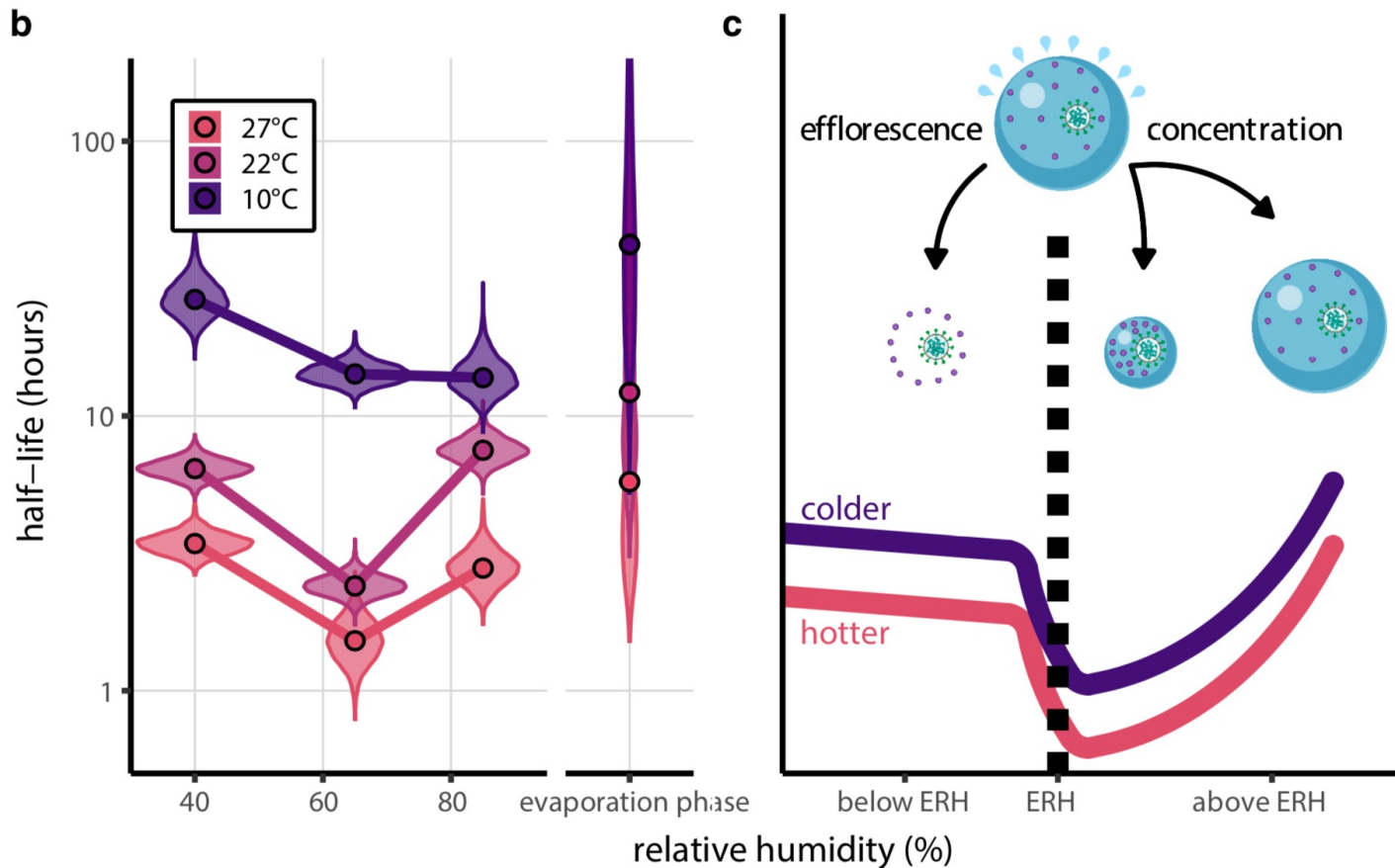
- ▶ **Study:** Testing flu viability in mucus, under a wide range of humidity
 - Including humidity near 100%
- ▶ **Result: Bimodal response**



Yang et al. *PLoS One* 2012

Laboratory studies: SARS-CoV-2/COVID-19

- ▶ Estimated half-life of SARS-CoV-2 on an inert surface vs. Relative Humidity & Temperature (Morries et al. 2021 *eLife*)

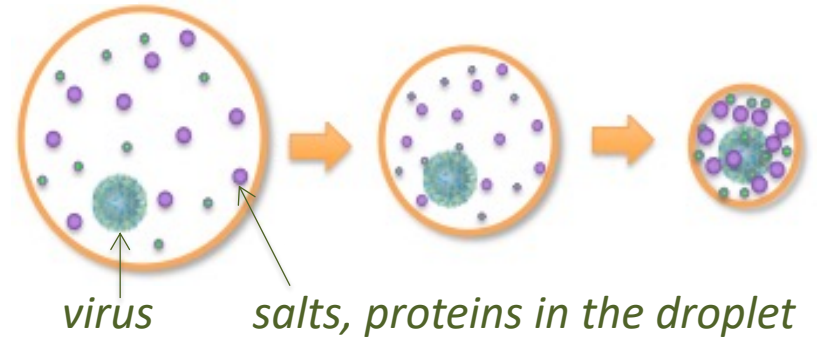
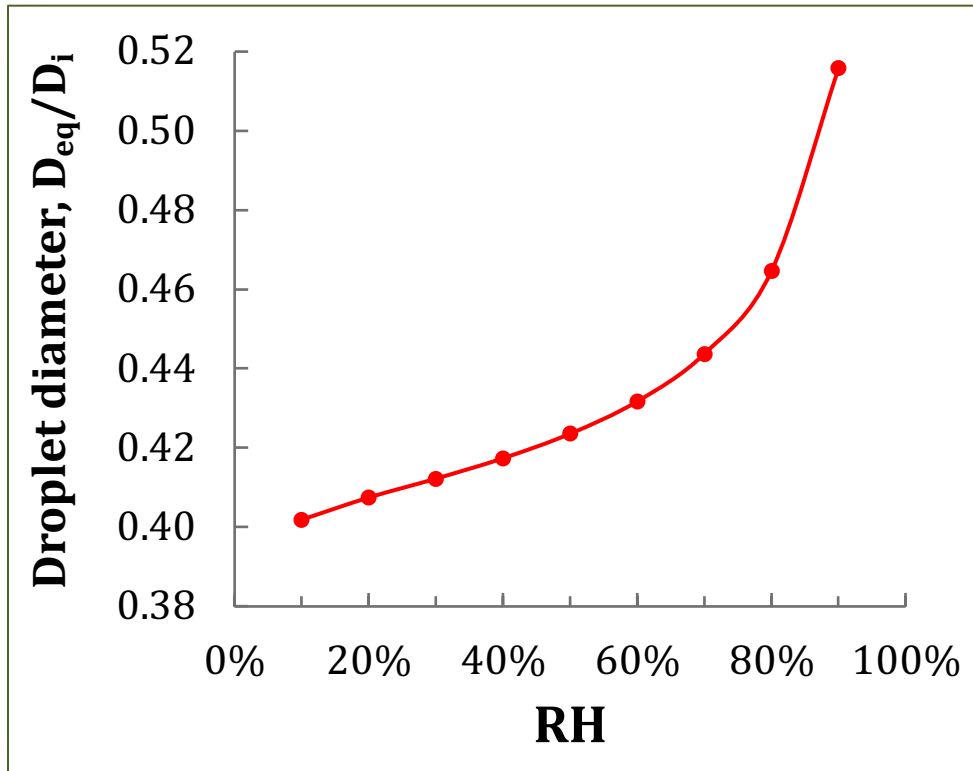


Morries et al. 2021 *eLife*

Potential mechanisms

► Relative humidity v. droplet size (Köhler theory)

$$RH = \frac{P_{\infty}}{P_{sat}} \overset{\text{Equilibrium}}{\underset{P_{\infty} = P_d}{=}} \frac{P_d}{P_{sat}} = \exp\left(-\frac{6imM_w}{M_s \rho \pi d^3}\right) \exp\left(\frac{4\gamma M_w}{\rho RTd}\right)$$



*Droplet shrinks by half -->
 Concentration increases by 8 times
 $D \rightarrow \frac{1}{2} D$
 Vol. --> $\frac{1}{8}$ Vol.
 Conc. --> 8 Conc.*

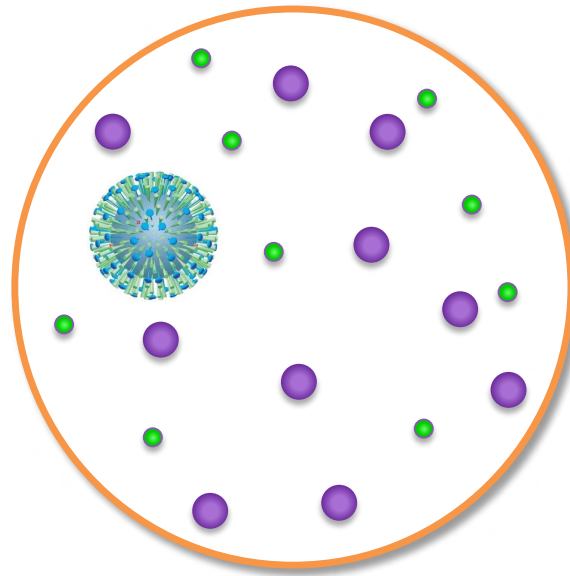
Humidity changes => The micro-environment surrounding the virus changes!

Potential mechanisms: Viability and Three RH Regimes

Low RH (<50%)

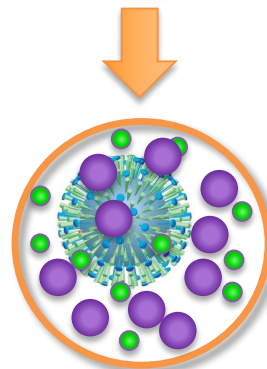
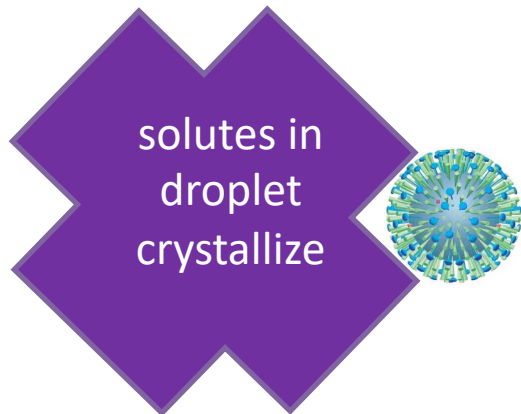
Efflorescence occurs
-> solutes cannot
harm virus ->
viability is
maintained

Medium RH



Very high RH

Minimal
evaporation ->
physiological
conditions are
maintained in
droplet -> viability is
maintained

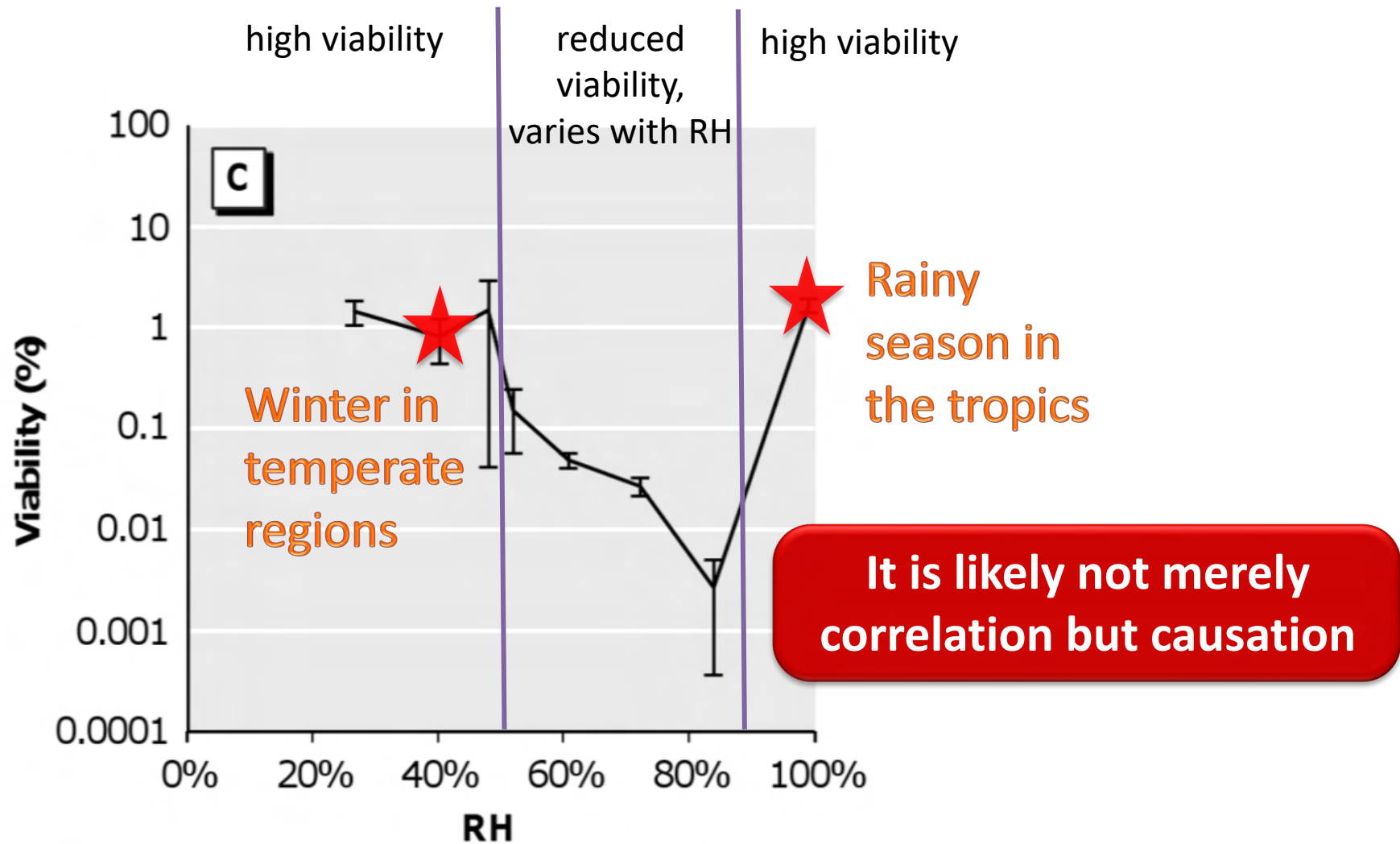


Elevated solute
concentrations may
harm virus

Slide credit: Linsey Marr; Yang et al., 2012, *PLoS One*; 2012 *AEM*

Potential mechanisms:

Viability, Three RH Regimes, and Seasonality



**MODEL:
MODELING THE SEASONALITY
USING CLIMATE/ENV. CONDITIONS**

Temperate regions: A humidity-forced SIRS model for influenza (low humidity -> high transmission)

▶ **The model:**

$$\frac{dS}{dt} = \frac{N - S - I}{L} - \frac{\beta(t)IS}{N}$$

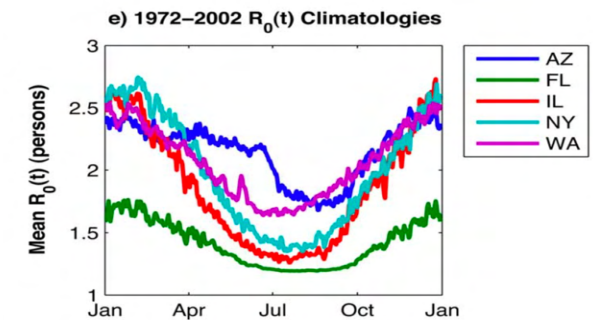
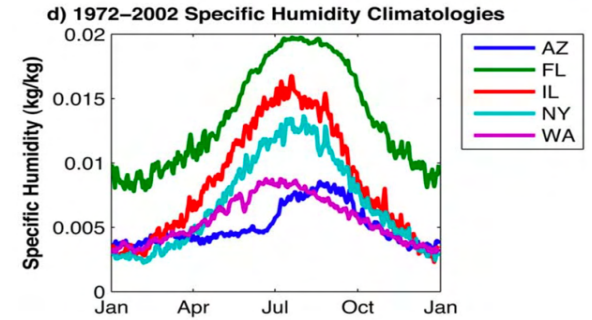
$$\frac{dI}{dt} = \frac{\beta(t)IS}{N} - \gamma I$$

○ $\beta(t)$ is modulated by humidity

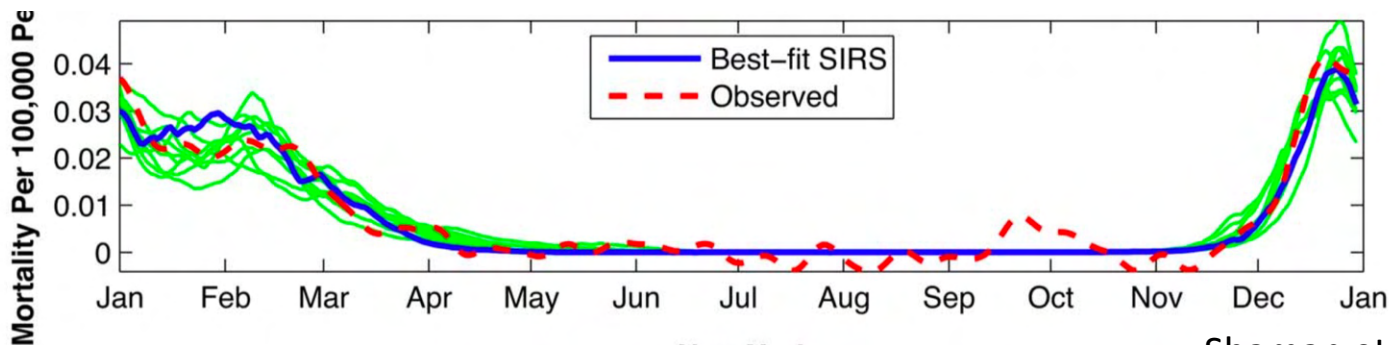
$$\beta(t) = (R_{0min} + (R_{0max} - R_{0min})e^{-180q(t)})\gamma$$

❖ $q(t)$: daily specific humidity

❖ R_{0min}/R_{0max} : min/max bound of R_0



▶ **Finding: The model is able to recreate the flu epidemic dynamics**

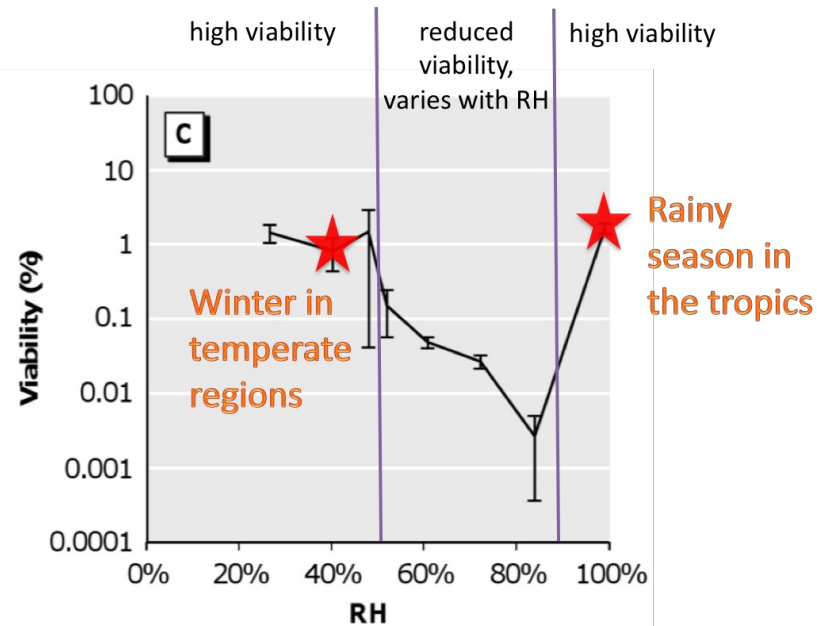


Shaman et al. 2010. PLoS Biol.

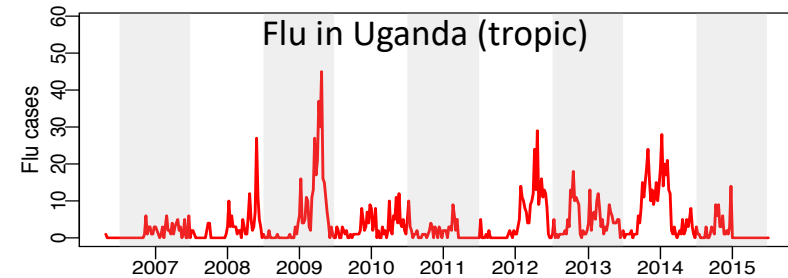
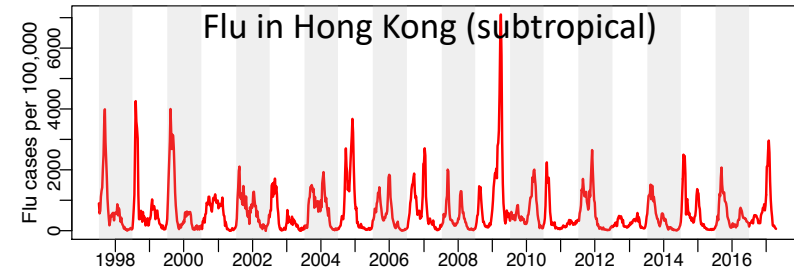
What about the (sub)tropics?

► Differences vs. temperate climates

- (Sub)tropics: Flu can occur year round
- Humidity: response is not monotonic
 - ❖ Recall: viability is high at both very low and very high humidity



- Temperature: Also play a key role



Yang et al. 2012 Plos One;
Yang et al. 2018a *Epidemics*;
Yang et al. 2018b *IRV*;
Yang et al. 2020 *PLoS Comput Biol*

A unified climate forced model (influenza)

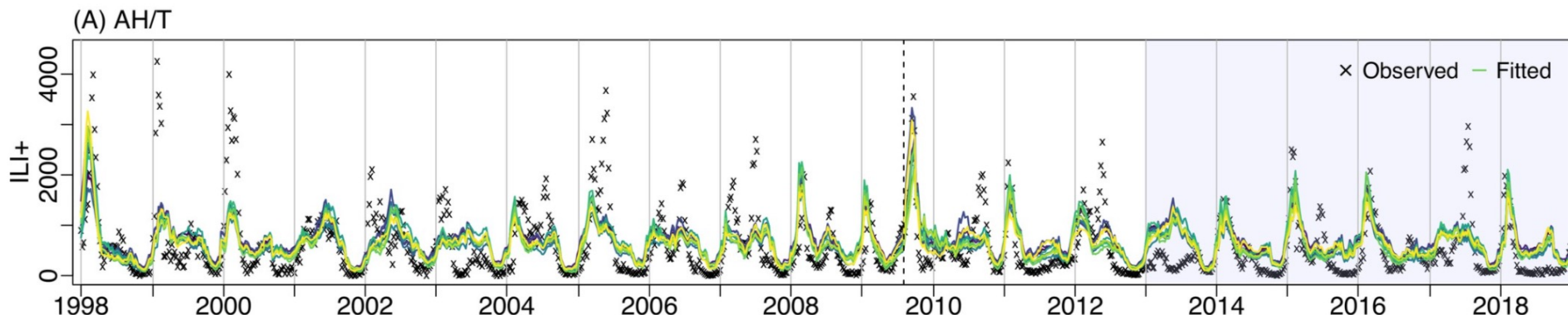
► **Model:**
$$R_0(t) = \underbrace{[aq^2(t) + bq(t) + c]}_{\text{Bimodal response to humidity, } q(t)} \underbrace{\left[\frac{T_c}{T(t)}\right]^{T_{exp}}}_{\text{Include the impact of temperature, } T(t)}$$

► **Model testing and parameter estimation:**

- Combined with an SIRS model, tested using flu incidence data in Hong Kong (subtropical) over ~20 years

► **Results:**

- Model captures the bimodal epidemics in HK



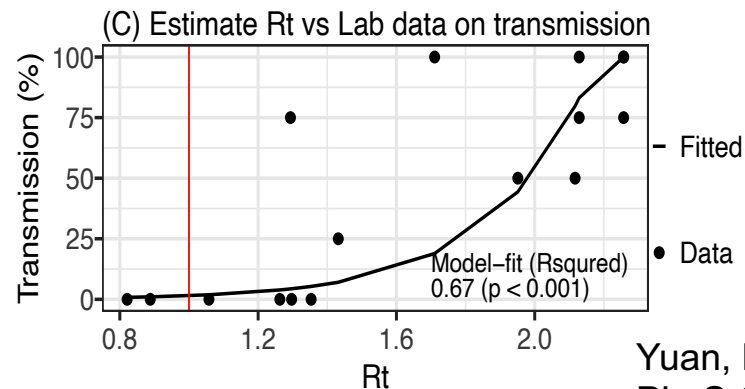
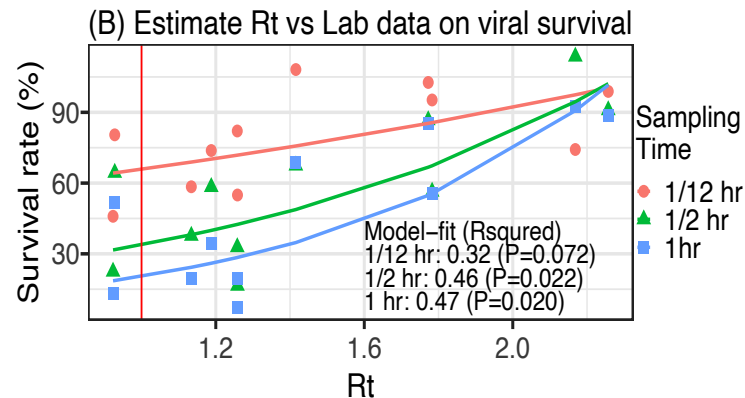
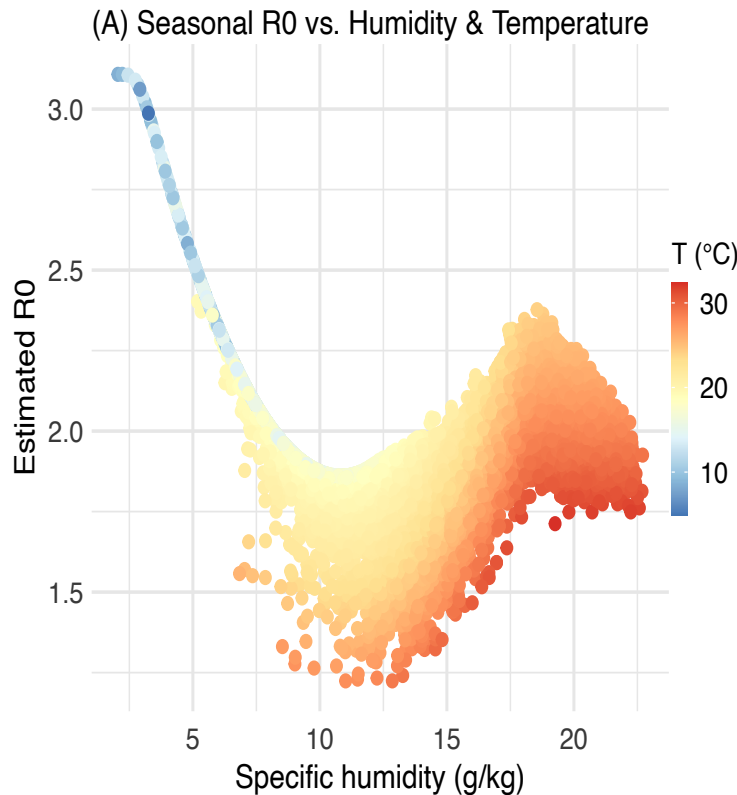
Yuan, Kramer et al. 2021 PLoS Comput Biol

A unified climate forced model (influenza)

► **Model form:** $R_0(t) = [aq^2(t) + bq(t) + c] \left[\frac{T_c}{T(t)} \right]^{T_{exp}}$

► **Results:**

- Model captures the bimodal epidemics in HK
- Covers both (sub)tropical and temperate climate conditions
- Consistent with lab virus survival/transmission data



Yuan, Kramer et al. 2021
PLoS Comput Biol

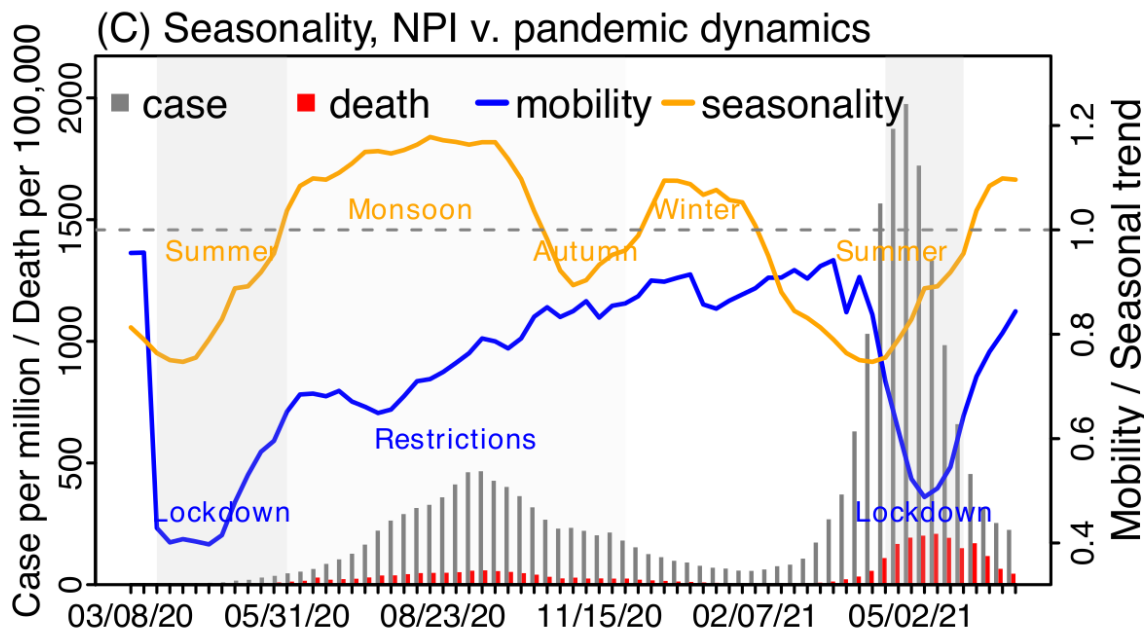
Can we use the model for SARS-CoV-2/COVID-19?

- ▶ **Method:** Scale the estimates to the annual average -> extract the seasonal trend
- ▶ **Incorporate the seasonal trend into epi-models to account for infection seasonality**

$$R_0(t) = [aq^2(t) + bq(t) + c] \left[\frac{T_c}{T(t)} \right]^{T_{exp}}$$

$$b_t = R_0(t) / \overline{R_0(t)}$$

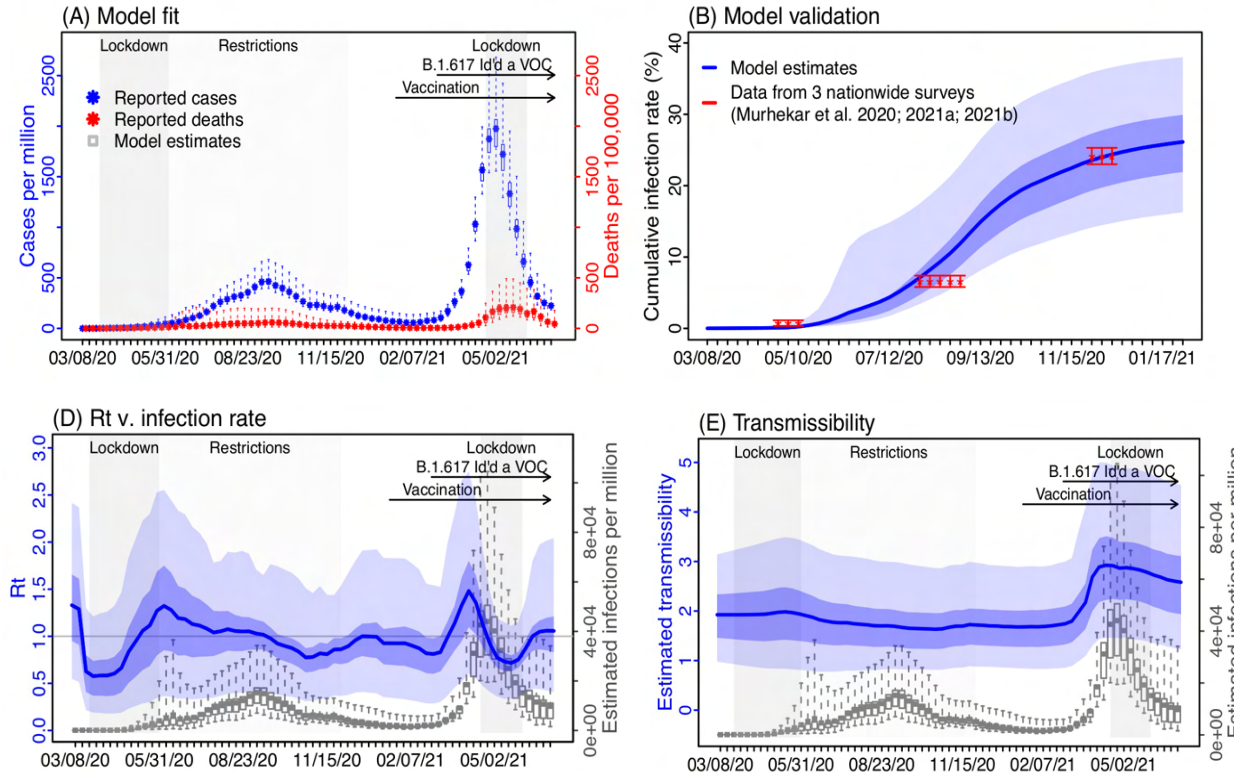
$$\begin{cases} \frac{dS}{dt} = \frac{R}{L_t} - \frac{b_t e_t m_t \beta_t IS}{N} - \varepsilon - v_{1,t} - v_{2,t} \\ \frac{dE}{dt} = \frac{b_t e_t m_t \beta_t IS}{N} - \frac{E}{Z_t} + \varepsilon \\ \frac{dI}{dt} = \frac{E}{Z_t} - \frac{I}{D_t} \\ \frac{dR}{dt} = \frac{I}{D_t} - \frac{R}{L_t} + v_{1,t} + v_{2,t} \end{cases}$$



Example Seasonal trend for India vs. other factors
(Yang & Shaman 2022 RSIF)

Example: Modeling COVID-19 in India (monsoon climate)

- ▶ **Study:** Combining key transmission determinants and data to model the COVID-19 pandemic in Indian (1st wave and the Delta wave)



Model validation showed the system is able to capture the underlying dynamics

R_t fluctuated over time
vs.
Transmissibility (R_{TX}) \nearrow
following rise of Delta

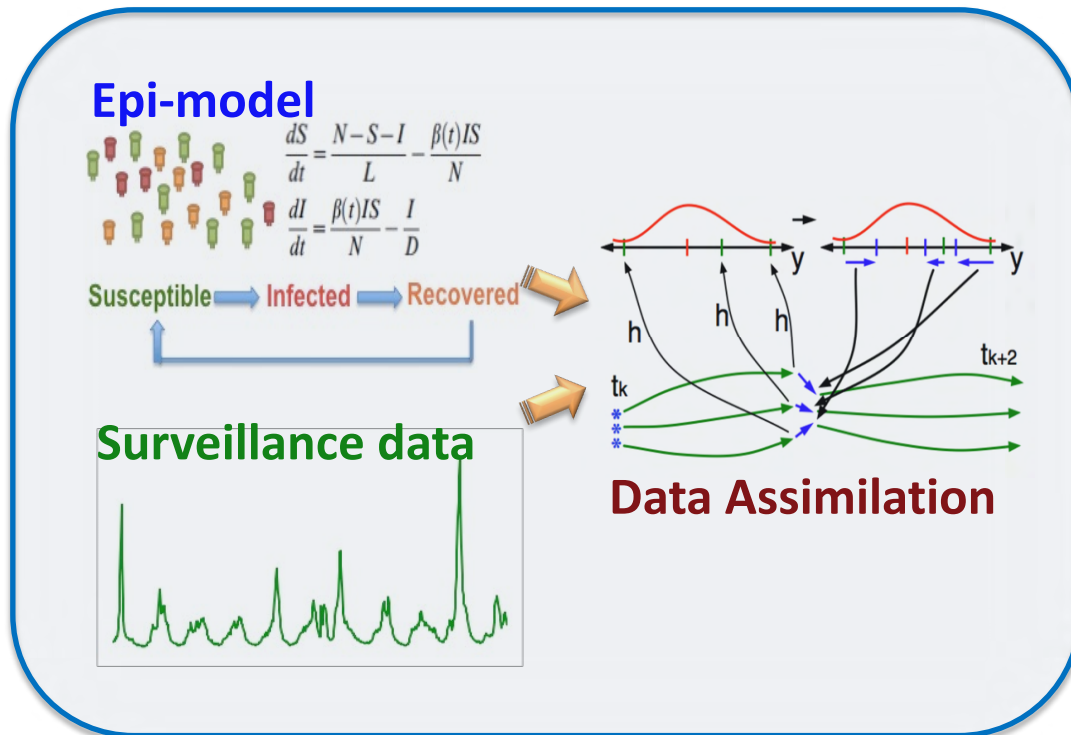
By accounting for seasonality, NPIs etc., we can **estimate variant-specific properties**
Applicable to diverse climate conditions and variants (Alpha in UK: NH temperate;
Beta/Omicron in S Africa: SH temperate; Gamma in Brazil: tropical; Delta in India: monsoon)

Yang & Shaman: 2021 Nature Comm; 2022 RSIF; 2022 eLife

**FORECAST:
USING THE SEASONALITY MODELS TO
IMPROVE FORECAST ACCURACY**

Overview of our forecast system

- ▶ 3 components: Model + Data + Data Assimilation
- ▶ 2 stages: Training + Forecast



- **Stage 1: Training**

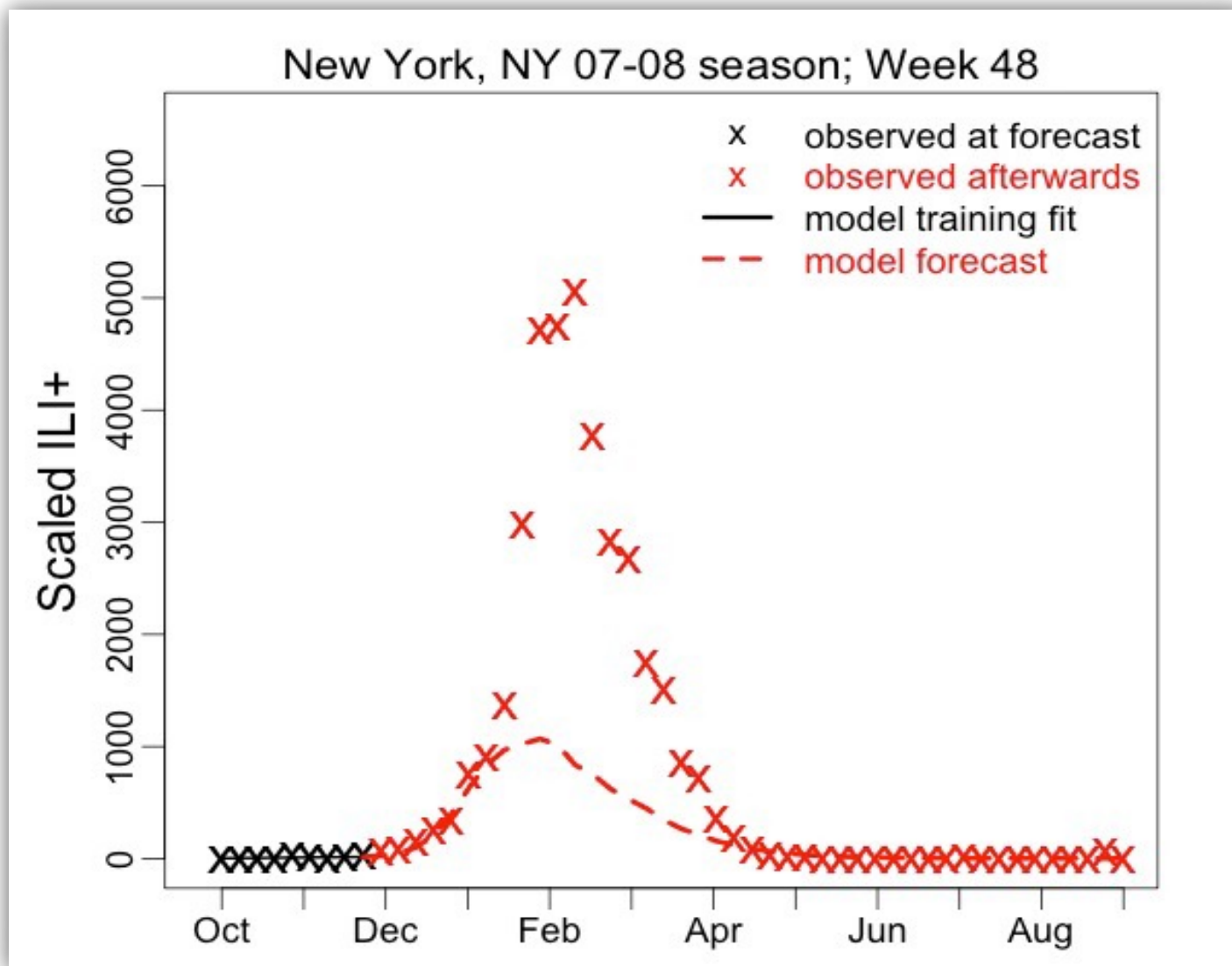
Optimize the model;
get initial conditions



- **Stage 2: Forecast**

Predict epi-unfolding

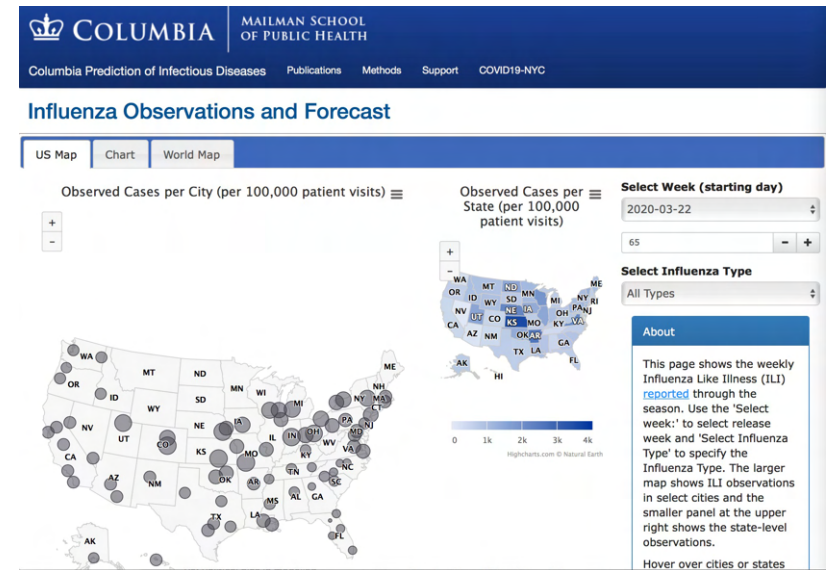
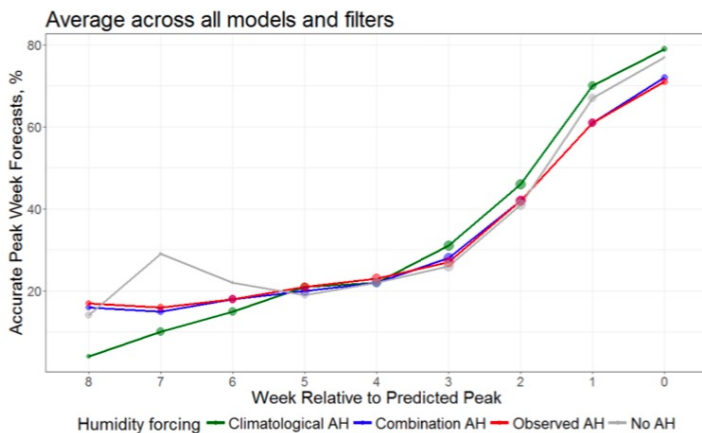
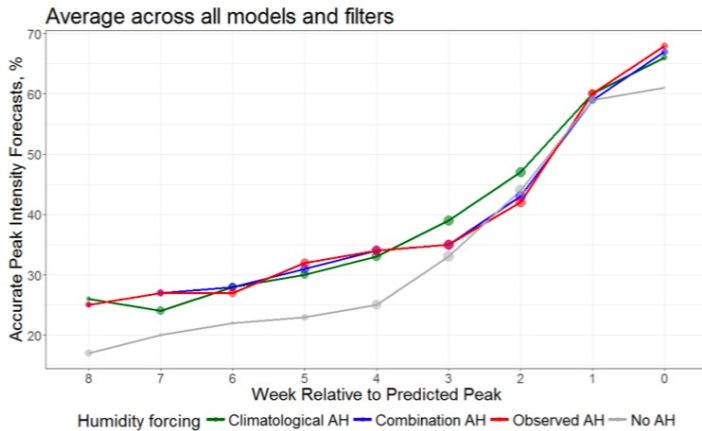
Example forecast



Example retrospective forecast. The forecast is increasingly accurate as more observations are assimilated.

Influenza forecast

- ▶ **Study:** Test the forecast accuracy using different settings of humidity forcing vs no humidity included in the model
- ▶ **Result:** The inclusion of humidity improves accuracy



The humidity-forced SIRS model has been used in flu real-time forecast since 2012
 Shaman et al.: <http://cpid.iri.columbia.edu>

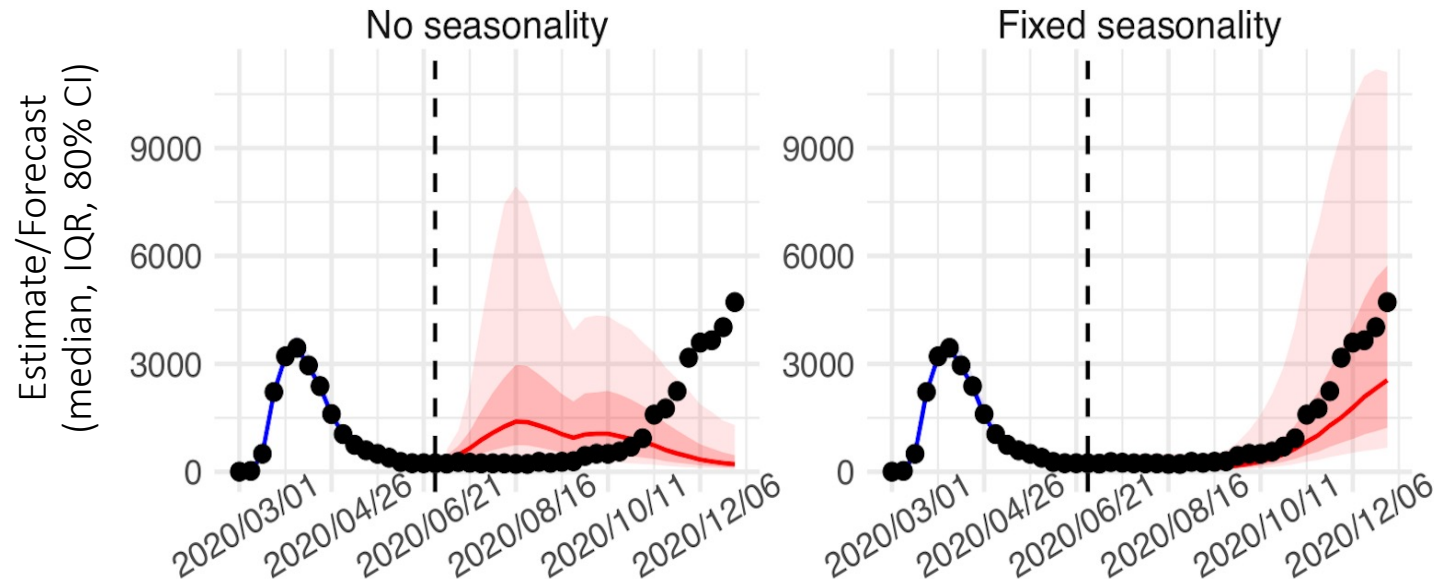
Fig 1. Percentage of forecasts accurate for predictions of peak intensity and peak timing

Shaman et al. 2017 PLoSCB

COVID-19 forecast

- ▶ **Study:** Test multiple strategies to improve long-lead COVID-19 forecast
 - Study period: July 2020 – Sep 2022 (multiple waves)
 - 10 states, one each from the 10 HHS regions -> different seasonality
- ▶ **Result:** Including seasonality -> high accuracy in general, more so during the respiratory virus season

(C) Example forecasts comparing seasonality settings
(New York; Cases; deflation = .9; new variants; 2020-07-05)



**PUBLIC HEALTH ACTION:
HOW WOULD BETTER UNDERSTAND
SEASONALITY HELP?**

How would this help public health?

- ▶ Use the seasonal timing/infection risk to guide vaccination campaign, public health messaging
- ▶ More accurate epidemiological parameter estimation -> better gauge risk
- ▶ More accurate forecast -> better aid public health preparation

▶ Example: Modeling and projection of COVID-19 in NYC

- Two versions of model-inference systems:
 - ❖ With vs. w/o seasonality
- The system with seasonality predicted the 2nd wave (Fall/winter 2020), 6 months in advance

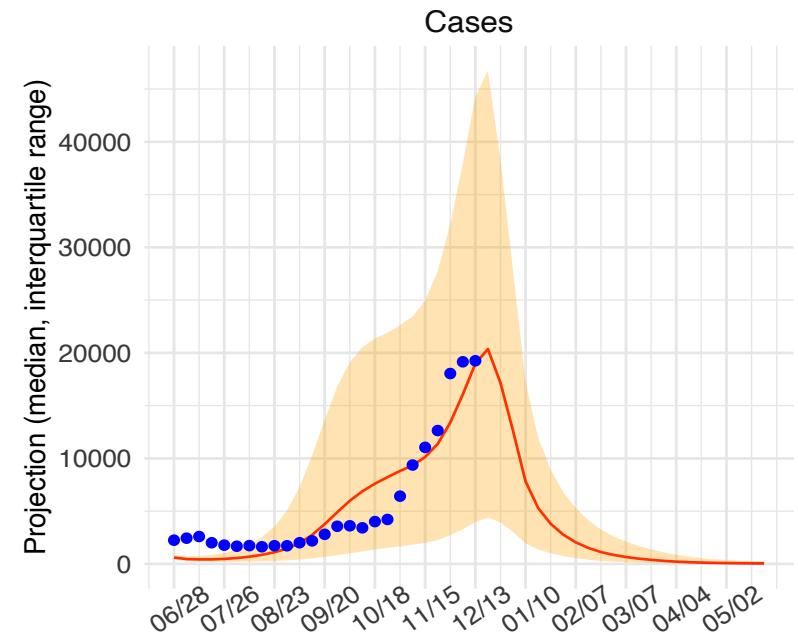


Fig: Model projection vs observations 6 months later
Red line: projection made on 6/30/2020;
surrounding area: projection interquartile range;
blue dots: observed (Yang et al. 2021 RSIF)

SUMMARY

Summary

- ▶ **Infection seasonality plays a key role in epidemic dynamics**
- ▶ **Climate/environmental conditions (humidity and temperature) can modulate infection seasonality**
 - This interaction is causal, mediated by several mechanisms
- ▶ **Incorporating the response to environmental factors can improve the accuracy of infectious disease models and infectious disease forecasts**
- ▶ **The model estimates and forecasts can be used to guide public health action**
- ▶ **“Knowledge is power”**

thank you!