

#### Flu Scenario Modeling Hub Report

13 December, 2022 Scenario Modeling Hub Team<sup>1</sup>

#### **Key Takeaways**

In Round 2, we assessed the impact of reduced residual immunity coming into the 2022-2023 influenza season as a result of reduced transmission during the last two influenza seasons because of the COVID-19 pandemic. We also assessed the impact of low versus high vaccine effectiveness. Ensemble projections are based on contributions from 11 teams (9 nationally) and cover the period from Nov 13, 2022 to June 3, 2023. Full scenario specifications can be found <a href="here">here</a>.

#### Key Takeaways from the Second Round

- Under the assumption of 50% reduced immunity at the start of the season, and as compared to the typical level of immunity pre-COVID-19 pandemic, a 40-50% higher hospitalization and death burden is expected based on median national projections.
- Increased vaccine effectiveness is expected to substantially decrease both the peak and cumulative hospitalizations regardless of existing population immunity, reducing hospitalization by 19.5% in low immunity scenarios and by 30.4% in high immunity scenarios.
- In the worst case scenario, with low vaccine effectiveness and low existing immunity, weekly hospitalizations are projected to peak at 38,000 hospitalizations nationally (50% 33,000-46,500). In this scenario, peak hospitalizations have a substantial likelihood to exceed the peak of the severe 2017-18 season. Observed hospitalizations and deaths are most closely tracking with the worst case scenario at the time of this report, though other scenarios remain plausible.
- Irrespective of scenario, ensemble hospitalizations are projected to peak in mid- to late-December, 2022. The worst case scenario of low prior immunity and low VE projects a peak in the week of December 24 nationally (50% PI December 10 December 31), though there is substantial variability between states.
- There is substantial agreement in the trajectory of individual models in this round.

#### A few caveats are worth noting:

- Some teams had issues calibrating the average pre-season immunity scenarios (A and C), sometimes leading to unrealistically high values of R0 to be consistent with the fast and early rise of the epidemic in the South-East. Further, some teams had trouble making projections for some of the smaller states that had a later epidemic start and lacked informative data for calibration. Recent observations indicate a sharp rise in these states in weeks 46-48. This led us to initiate a third round of influenza projections that will include calibration data until week 48 and will be released later in December.
- The amount of calibration data available for the new HHS influenza dataset remains limited, and testing
  practices could change between and within seasons. Similarly, the amount of death data available for
  calibration is limited.

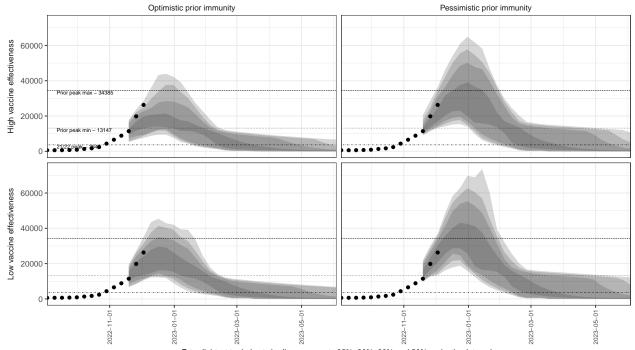
<sup>&</sup>lt;sup>1</sup>Compiled by Sara Loo, Shaun Truelove, Cecile Viboud, Lucie Contamin.

# Round 2 Scenario Specifications

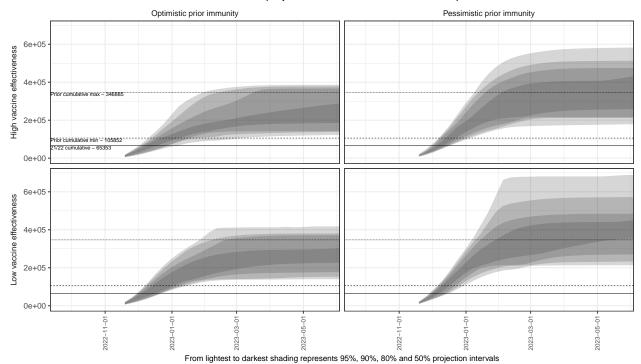
	Optimistic flu prior immunity  No impact of missed flu seasons due to the COVID-19 pandemic on prior immunity.*  Same amount of prior immunity as in a typical, pre-COVID19 pandemic prior season	Pessimistic flu prior immunity  Substantial impact of missed flu seasons due to the COVID-19 pandemic on prior immunity.*  50% lower immunity than a typical, pre-COVID19 pandemic season
High Vaccine Effectiveness  VE = 50% against medically attended influenza illnesses and hospitalizations (comparable to 2015-16 season)	Scenario A	Scenario B
Low Vaccine Effectiveness  ■ VE = 30% against medically attended influenza illnesses and hospitalizations (comparable to 2018-19 season)	Scenario C	Scenario D

## Ensemble projection intervals

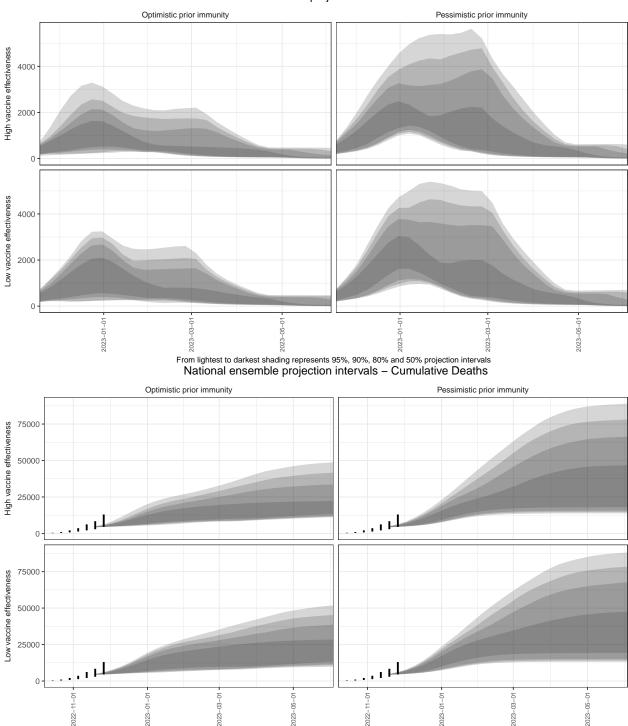
#### National ensemble projection intervals - Hospitalizations



From lightest to darkest shading represents 95%, 90%, 80% and 50% projection intervals National ensemble projection intervals – Cumulative Hospitalizations



#### National ensemble projection intervals - Deaths



Horizontal lines are given for prior peak incident and cumulative hospitalizations, from seasons from 2012-13 to 2019-20. The minimum and maximum peaks across these seasons are taken from FluSurv-NET (which is used as a proxy for hospitalizations). Nationally, the highest value is from the 2017-18 season, and the lowest from 2015-16. The 2021-22 flu season based on HHS data is also included to mark a small season.

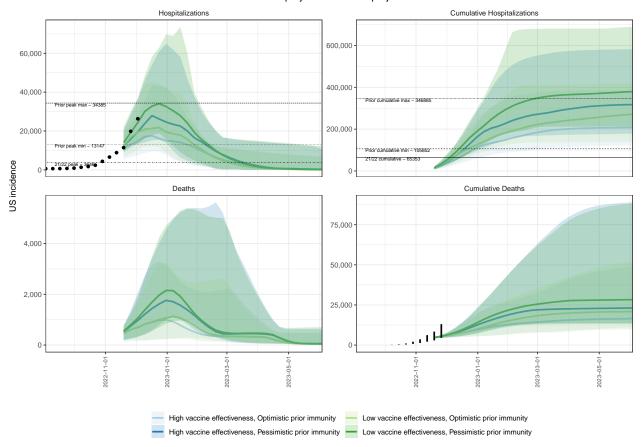
From lightest to darkest shading represents 95%, 90%, 80% and 50% projection intervals

Cumulative death observations are visualized as vertical bars, and are based on in-season death estimates provided by the CDC multiplier approach, which takes into account in-hospital flu deaths reported to

FluSurv-NET, propensity to test for flu in the hospital, and risk of dying of flu outside of the hospital. Bars represent 95% uncertainty intervals. Cumulative estimates are shown since the start of the flu season on Oct  $1,\,2022.$ 

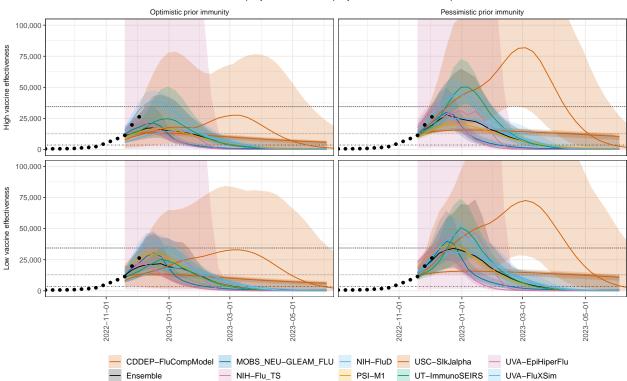
## National ensemble projections

Ensemble projections for national incident and cumulative hospitalizations and deaths separated by scenario. US ensemble projections & 95% projection intervals



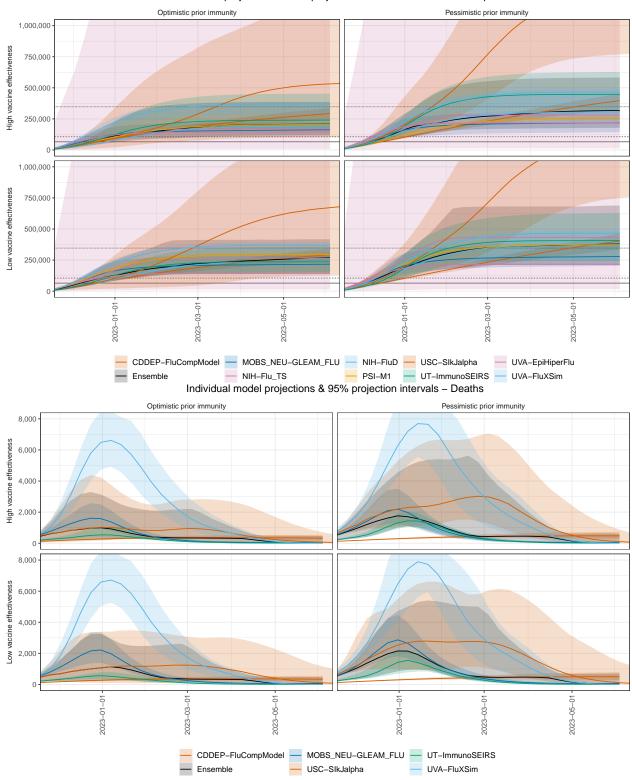
#### National individual model projections

Individual model projections and ensemble by scenario for national hospitalizations, deaths and cumulative hospitalizations. For visualization we set axes limits; full confidence intervals are shown as a supplemental plot on page 17-18.

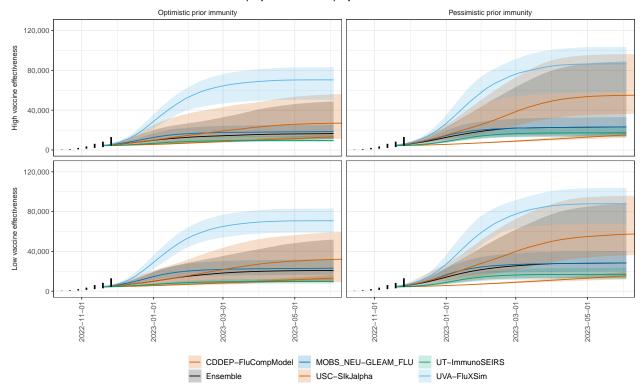


Individual model projections & 95% projection intervals - Hospitalizations

#### Individual model projections & 95% projection intervals – Cumulative Hospitalizations

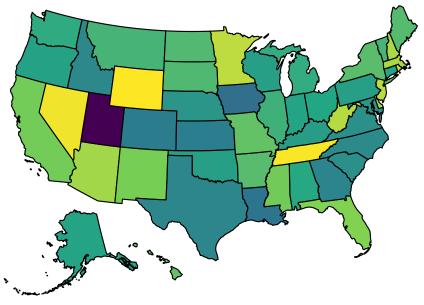


#### Individual model projections & 95% projection intervals – Cumulative Deaths



#### Risk maps

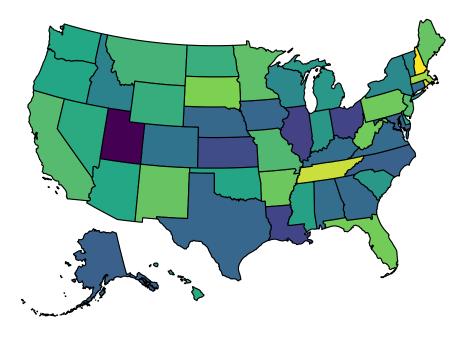
Peak incident reported hospitalizations per 10,000 population in scenario with low vaccination, and pessimistic immunity: November 13, 2022 to June 03, 2023



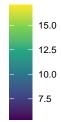
Peak incident hosps per 10,000 pop



Cumulative reported hospitalizations per 10,000 population in scenario with low vaccination, and pessimistic immunity: November 13, 2022 to June 03, 2023



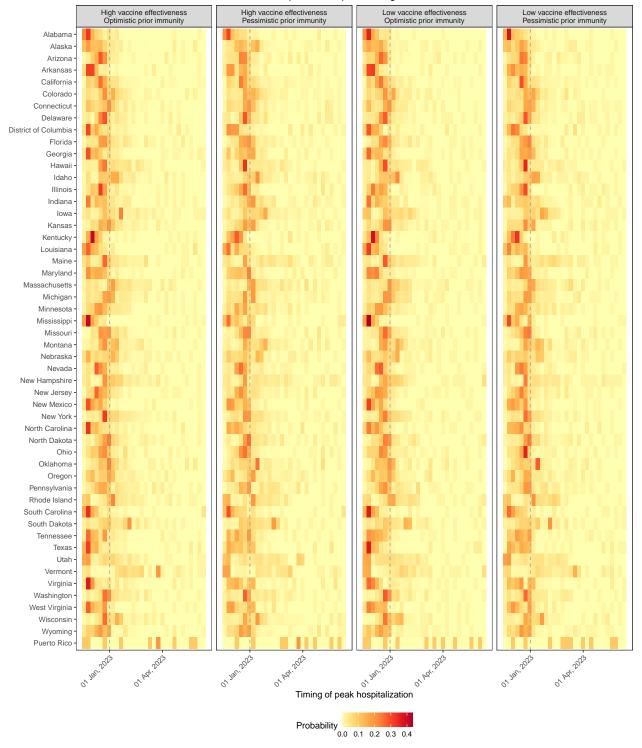
Cumulative hosps per 10,000 pop



#### State variability in peak timing

Ensembles projections for state-level timing of peak hospitalization incidence.

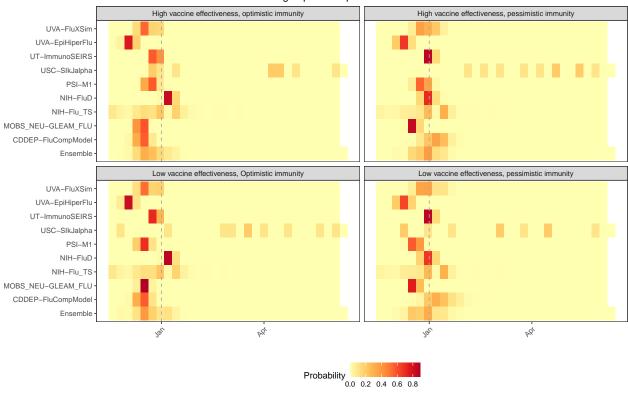
#### Hospitalization peak timing variation



#### Peak hospitalizations timing

Individual model probabilities for national timing of peak hospitalizations.

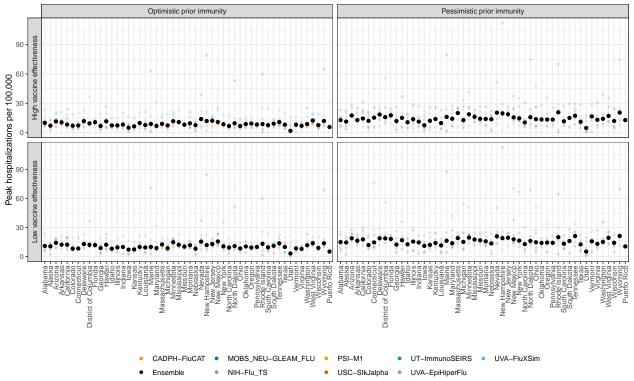
#### Timing of peak hospitalization across models



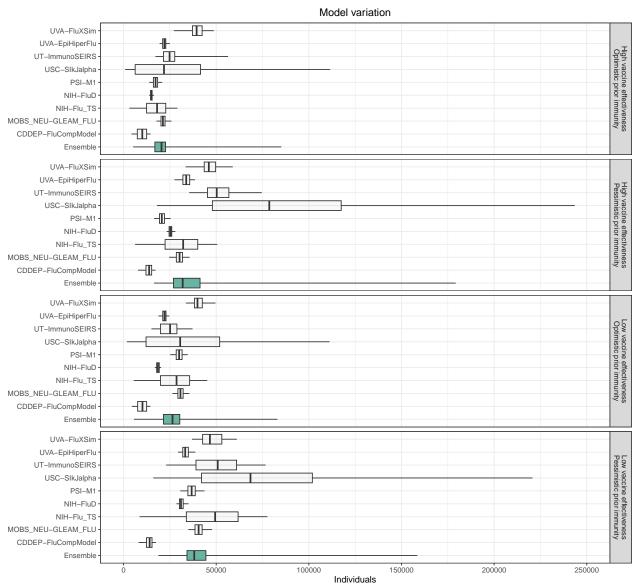
## State-level deviation in hospitalization incidence

Individual model and ensembles projections for state-level peak hospitalization incidence.

#### State variation in peak hospitalizations per 100,000



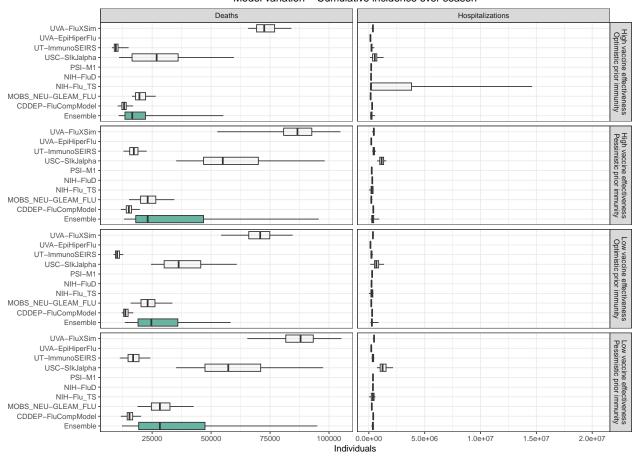
#### Model Variation in National Peak Size



Outliers are removed from boxplots for visualization.

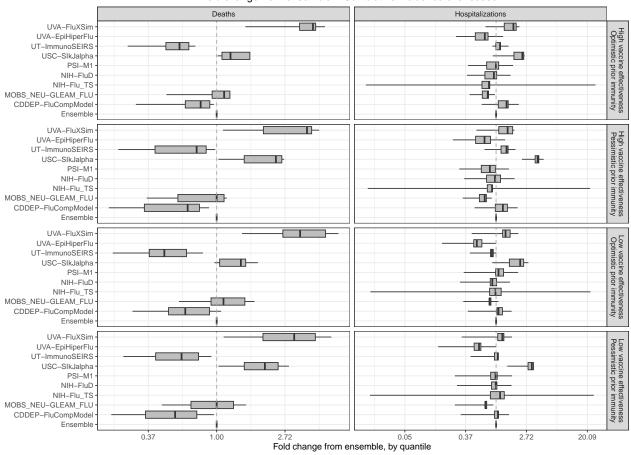
#### Cumulative incidence over season by model

#### Model variation - Cumulative incidence over season



#### Difference between model and ensemble distributions

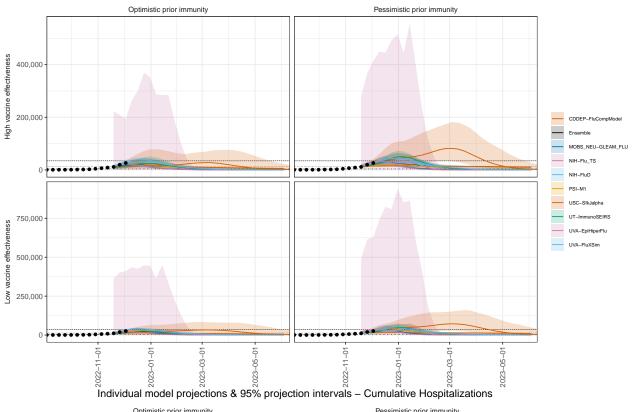


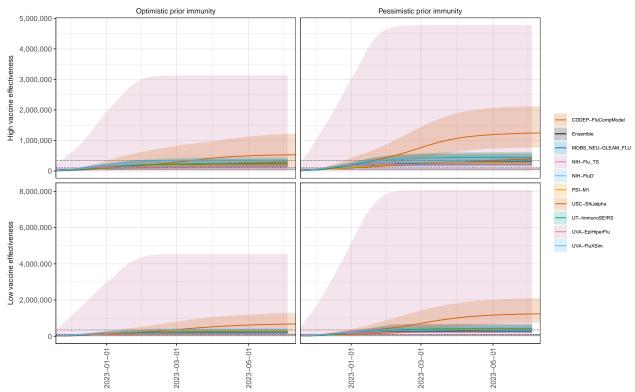


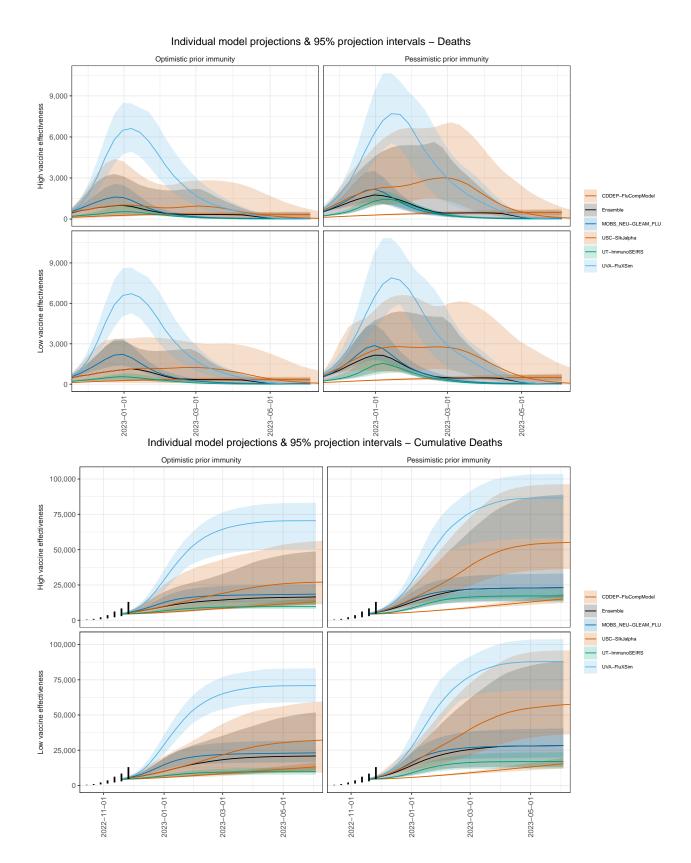
## Supplemental Plots

## National individual model projections - full confidence intervals

Individual model projections & 95% projection intervals – Hospitalizations



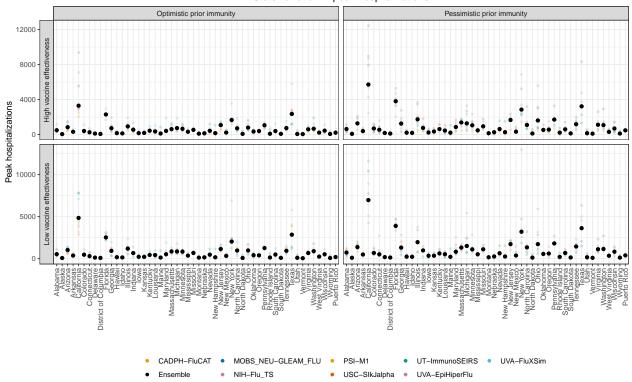




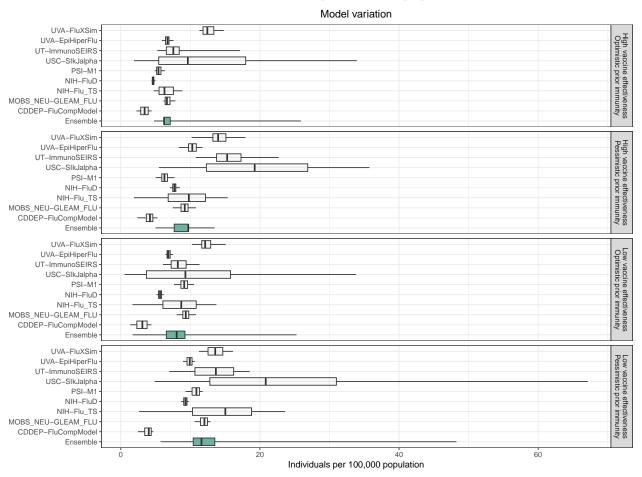
#### State-level deviation in hospitalization incidence

Individual model and ensembles projections for state-level peak hospitalization incidence.

#### State variation in peak hospitalizations

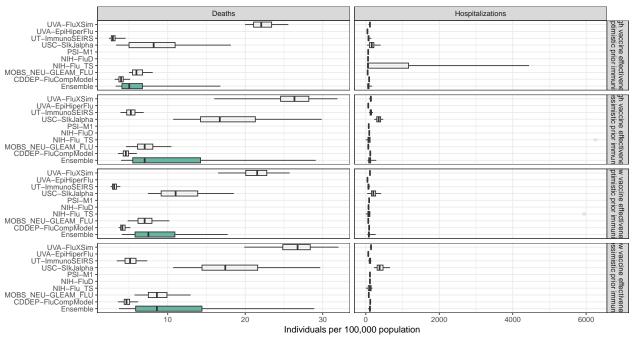


## Model Variation in National Peak Size - rates per 100,000 population



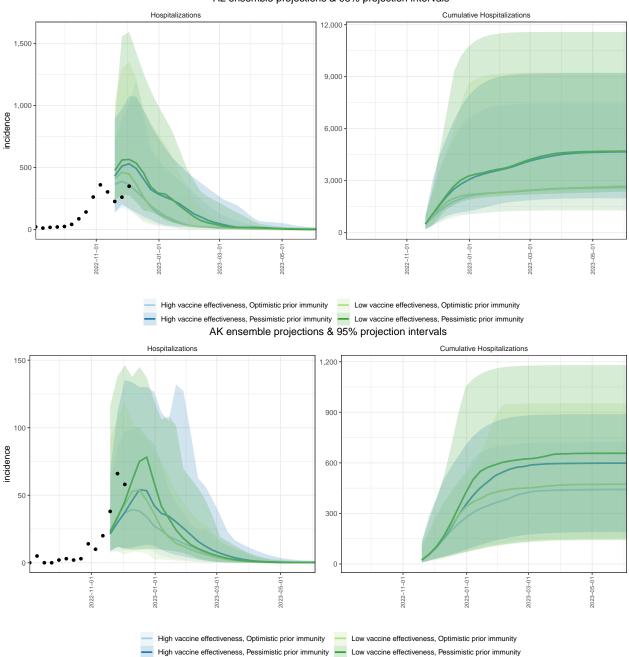
#### Cumulative incidence over season - rates per 100,000 population

#### Model variation - Cumulative incidence over season

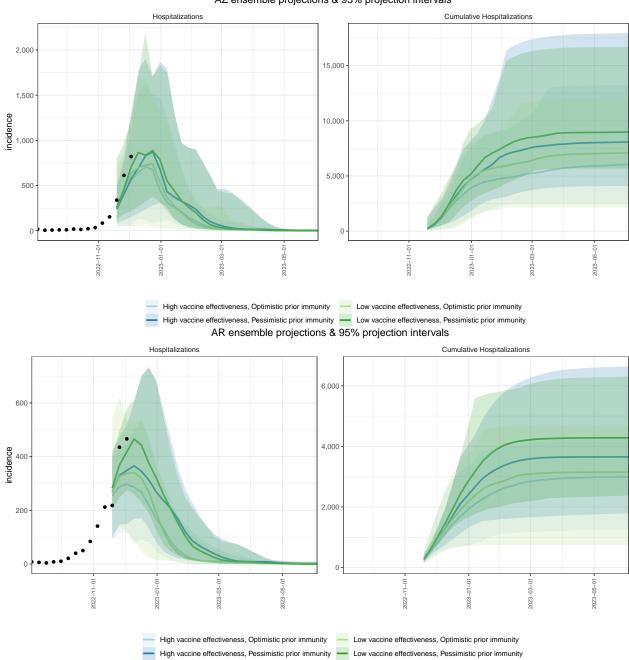


## State-level ensemble plots

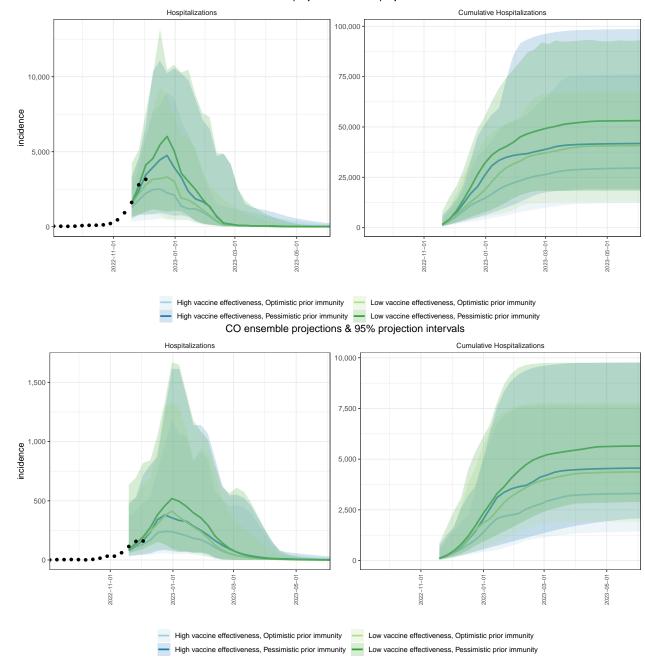
## AL ensemble projections & 95% projection intervals



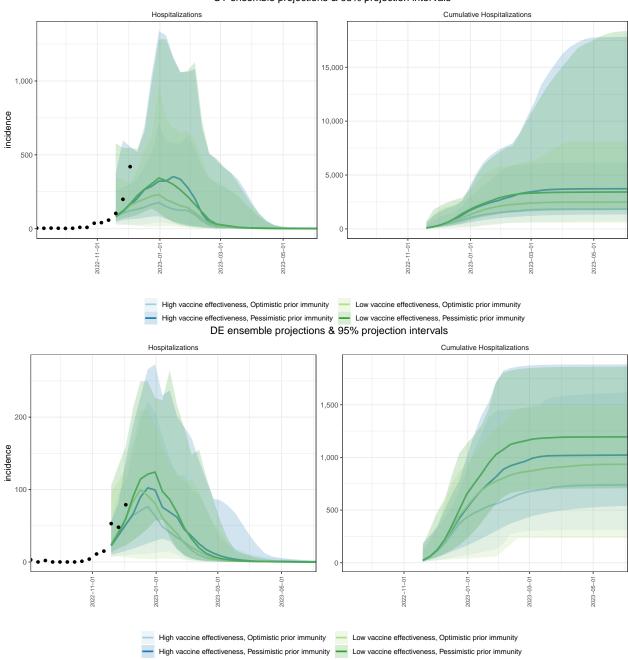
#### AZ ensemble projections & 95% projection intervals



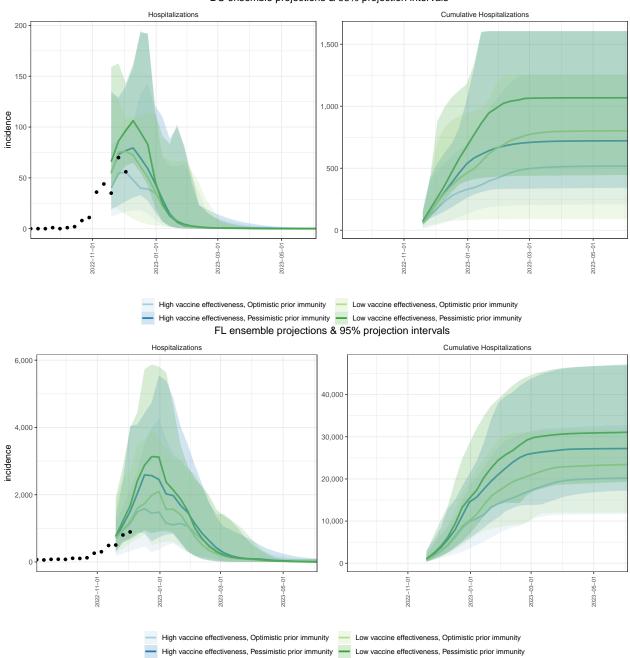
#### CA ensemble projections & 95% projection intervals



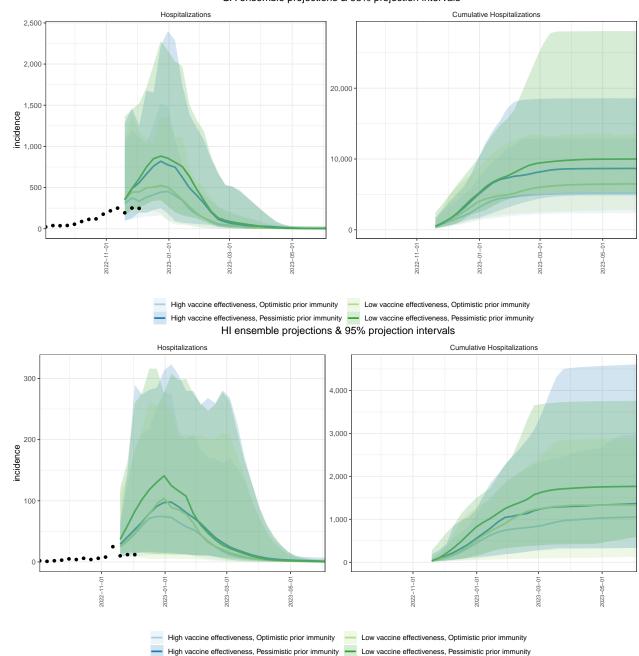
#### CT ensemble projections & 95% projection intervals



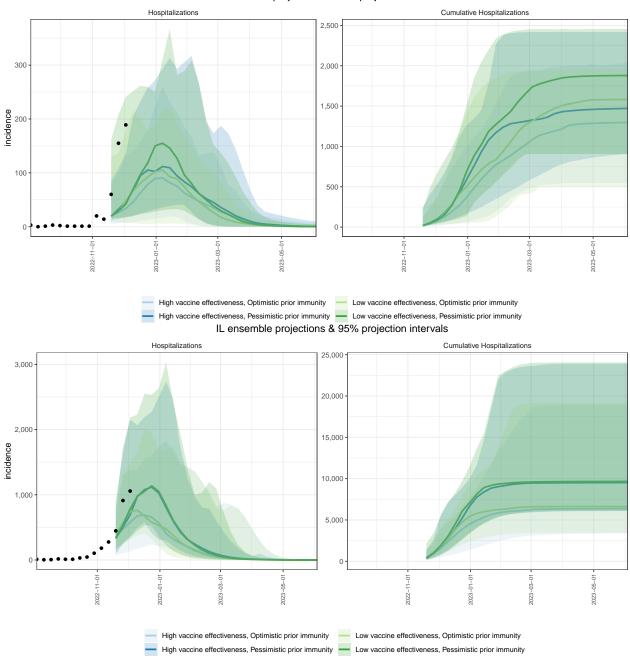
#### DC ensemble projections & 95% projection intervals



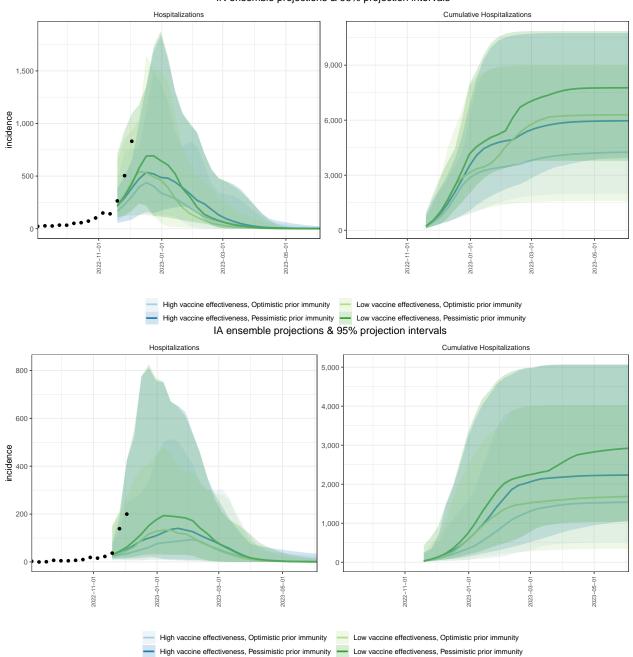
#### GA ensemble projections & 95% projection intervals



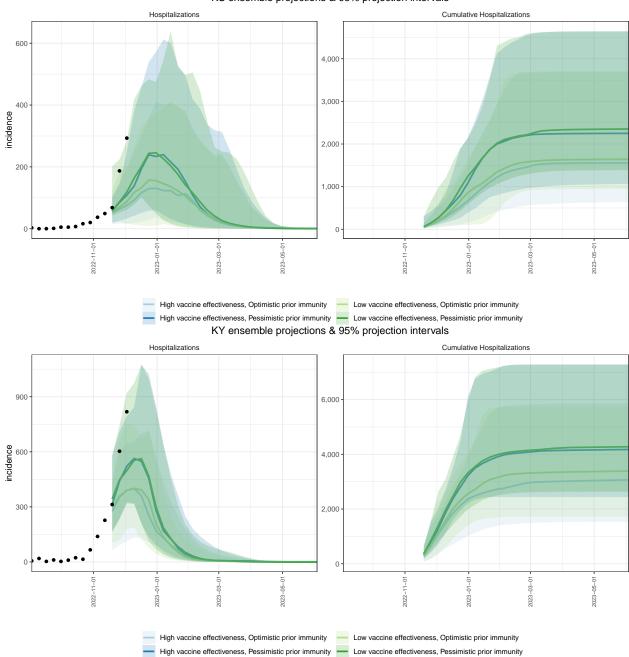
#### ID ensemble projections & 95% projection intervals



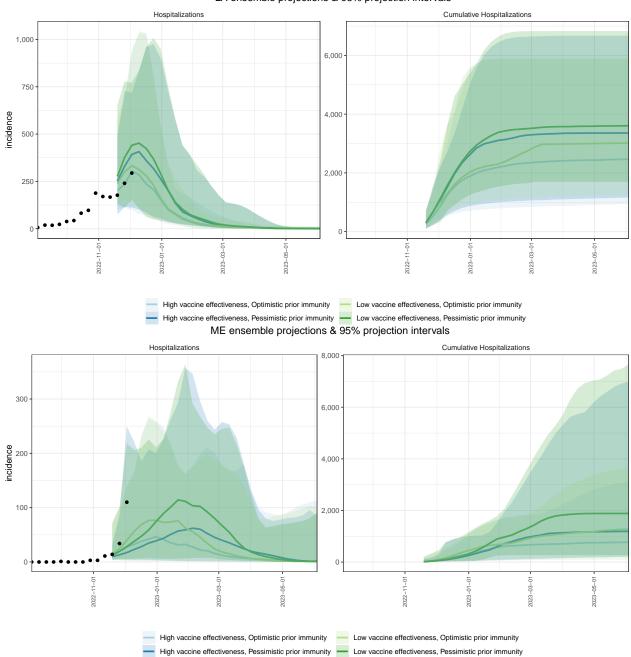
#### IN ensemble projections & 95% projection intervals



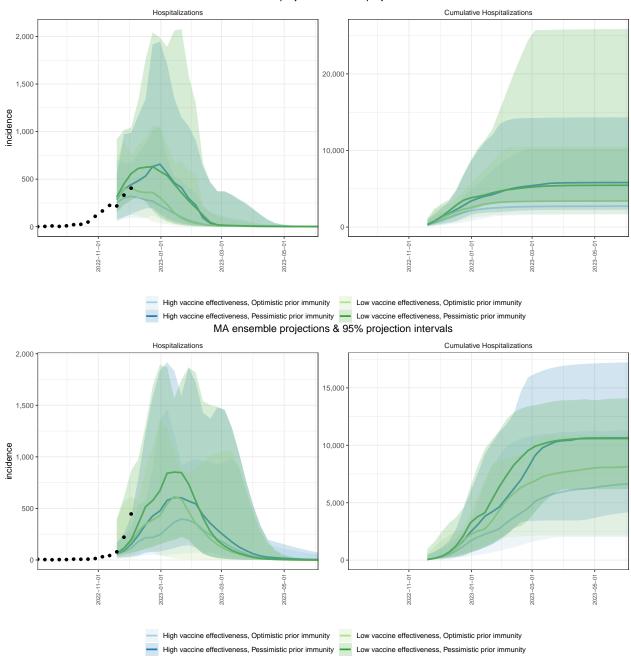
#### KS ensemble projections & 95% projection intervals



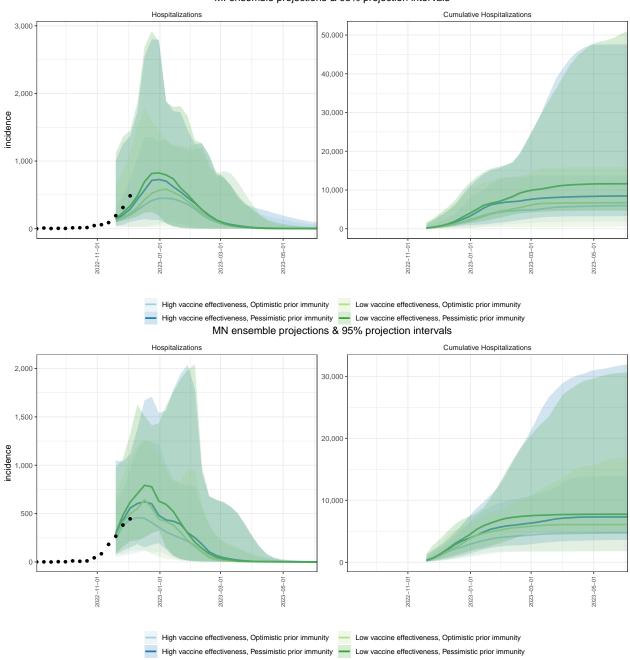
#### LA ensemble projections & 95% projection intervals



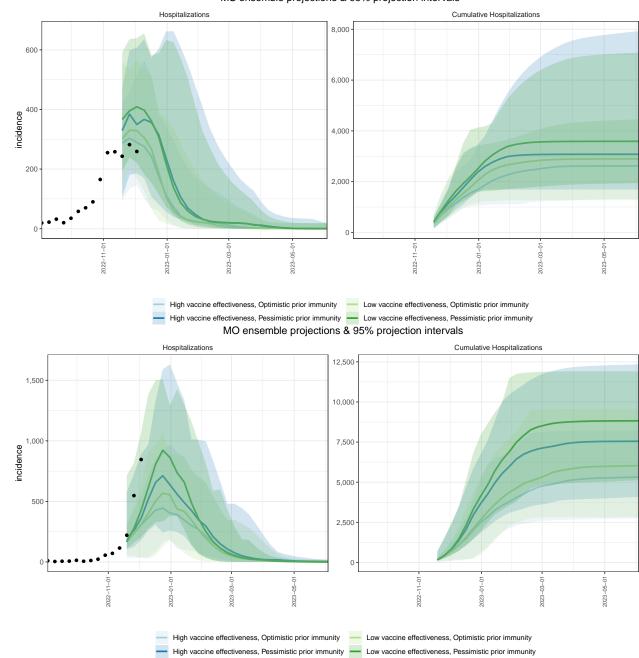
#### MD ensemble projections & 95% projection intervals



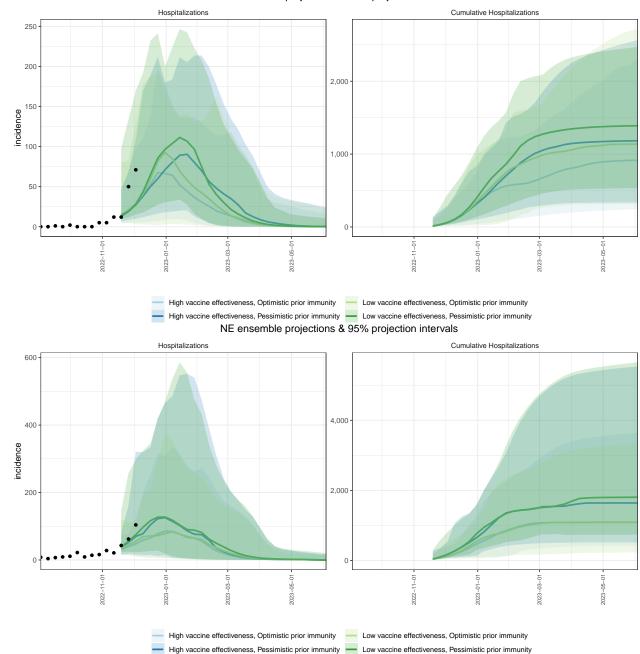
#### MI ensemble projections & 95% projection intervals



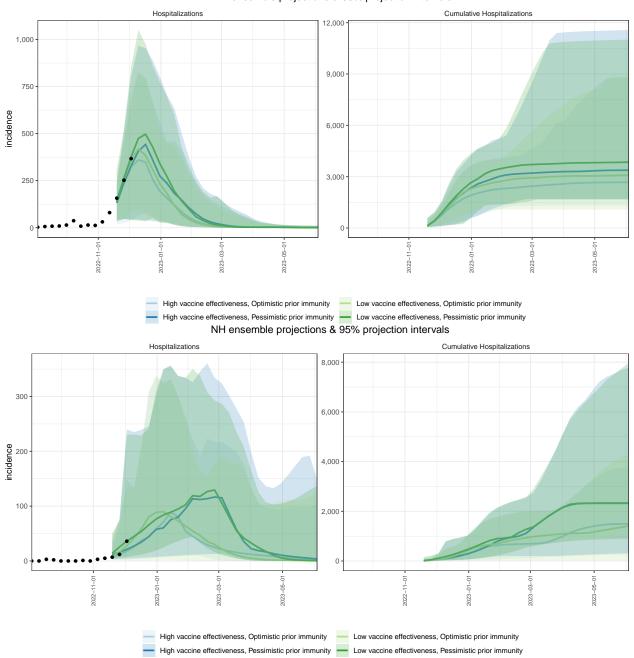
#### MS ensemble projections & 95% projection intervals



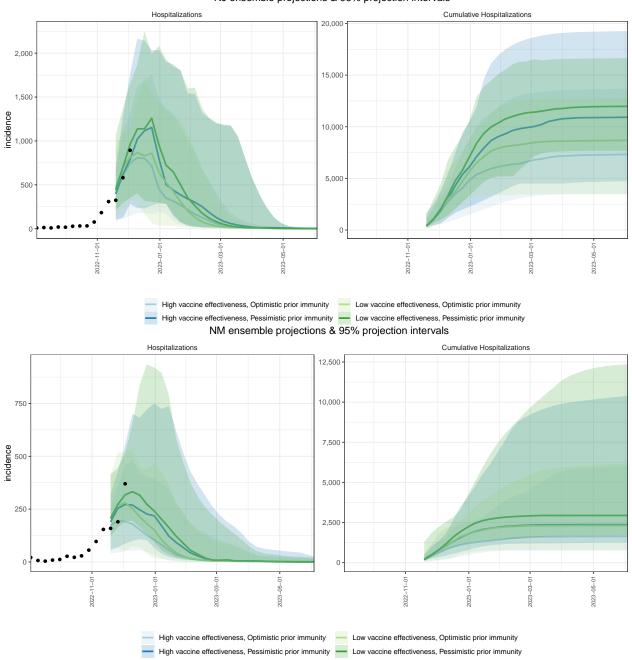
#### MT ensemble projections & 95% projection intervals



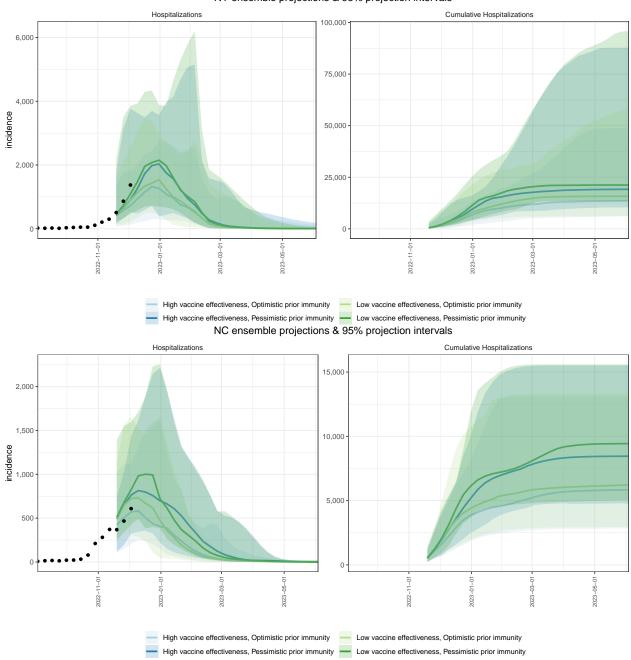
#### NV ensemble projections & 95% projection intervals



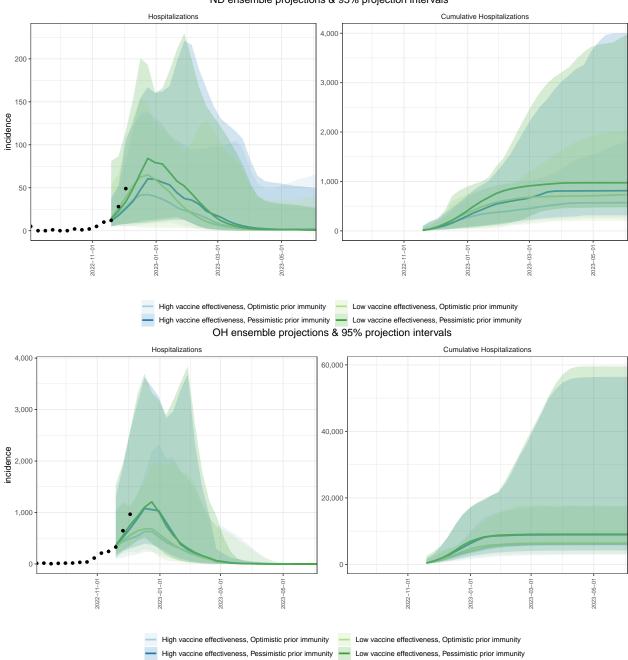
# NJ ensemble projections & 95% projection intervals



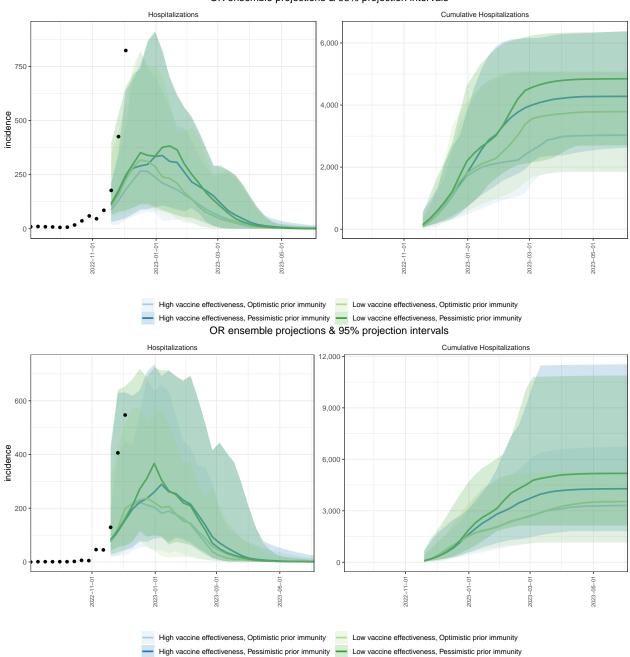
# NY ensemble projections & 95% projection intervals



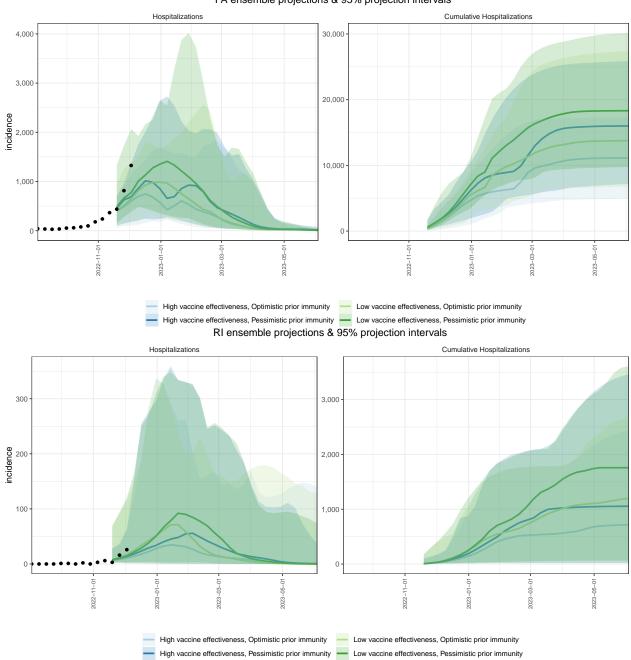
# ND ensemble projections & 95% projection intervals



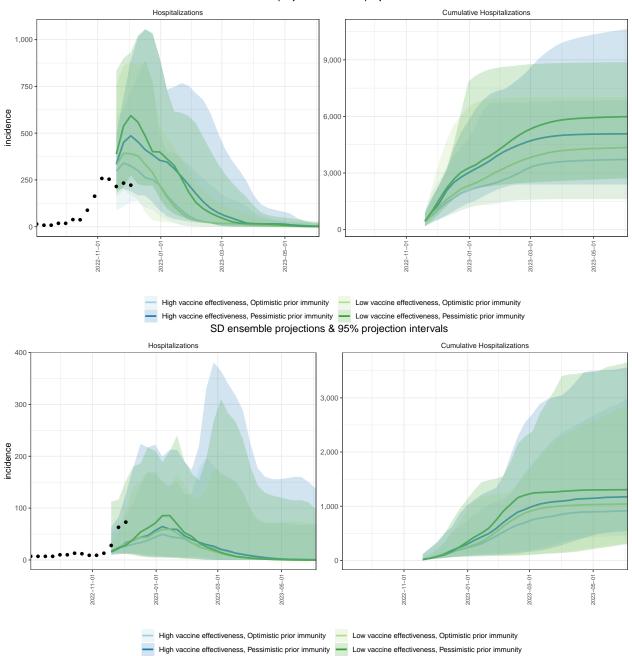
# OK ensemble projections & 95% projection intervals



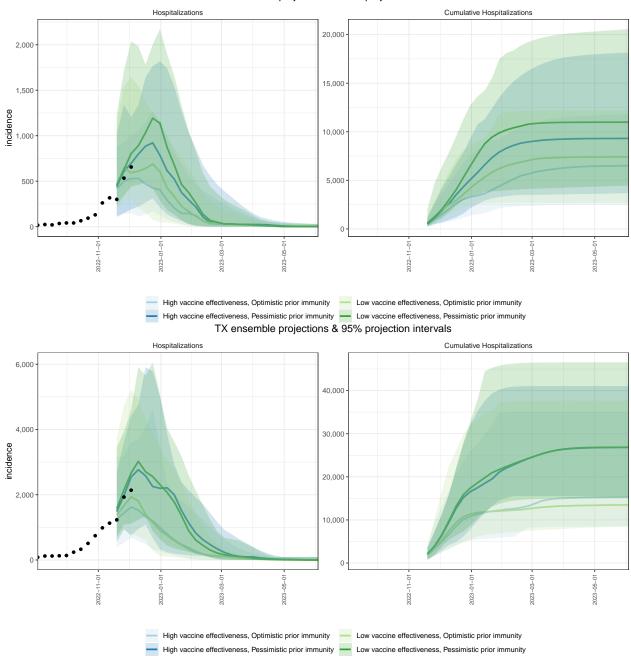
# PA ensemble projections & 95% projection intervals



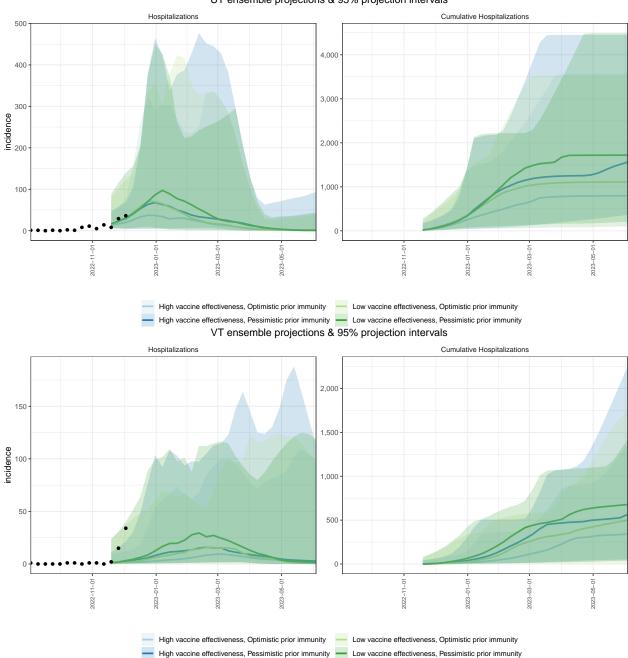
# SC ensemble projections & 95% projection intervals



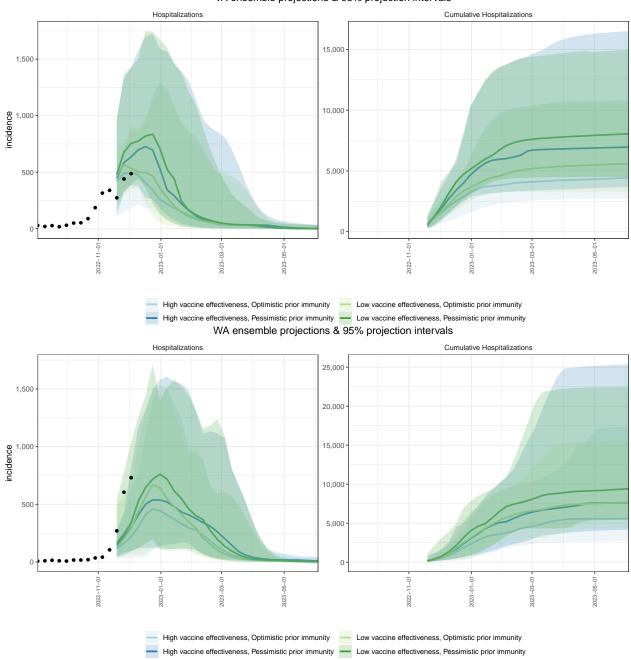
# TN ensemble projections & 95% projection intervals



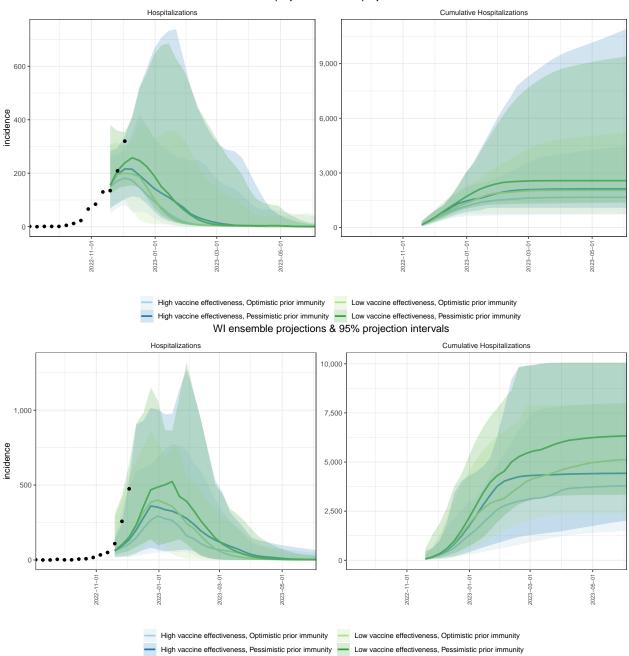
# UT ensemble projections & 95% projection intervals



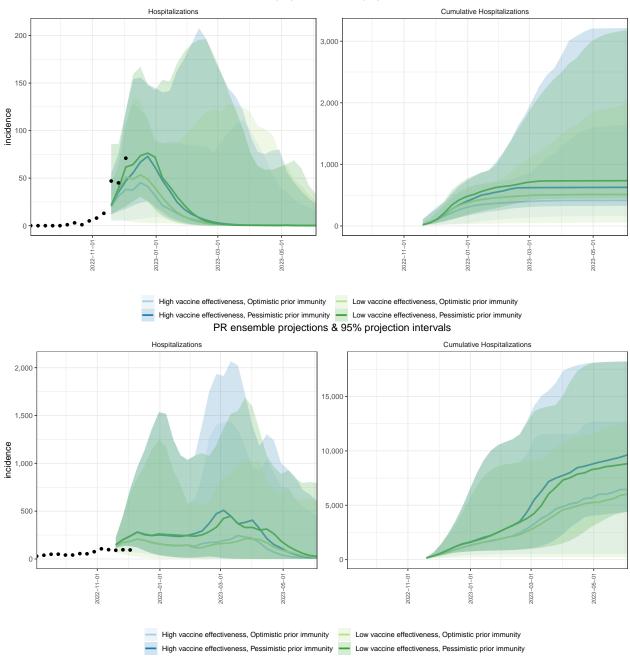
# VA ensemble projections & 95% projection intervals



# WV ensemble projections & 95% projection intervals

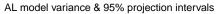


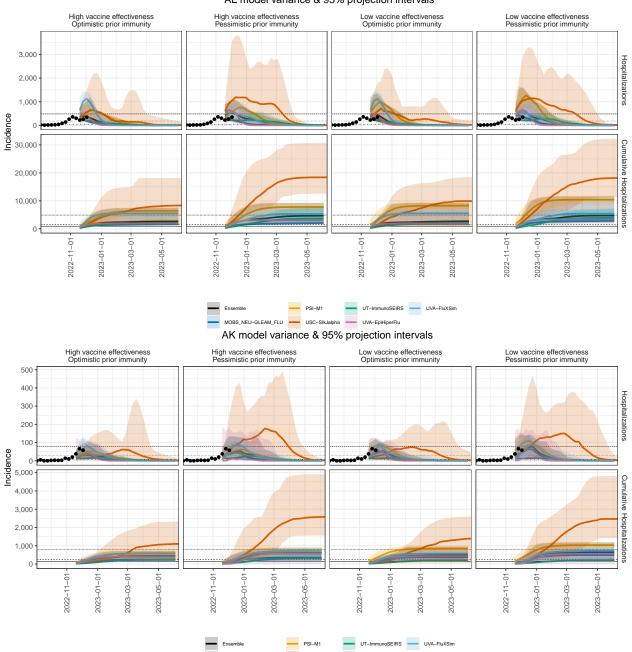
# WY ensemble projections & 95% projection intervals



# State-level model variation

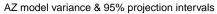
National model variation for all scenarios.

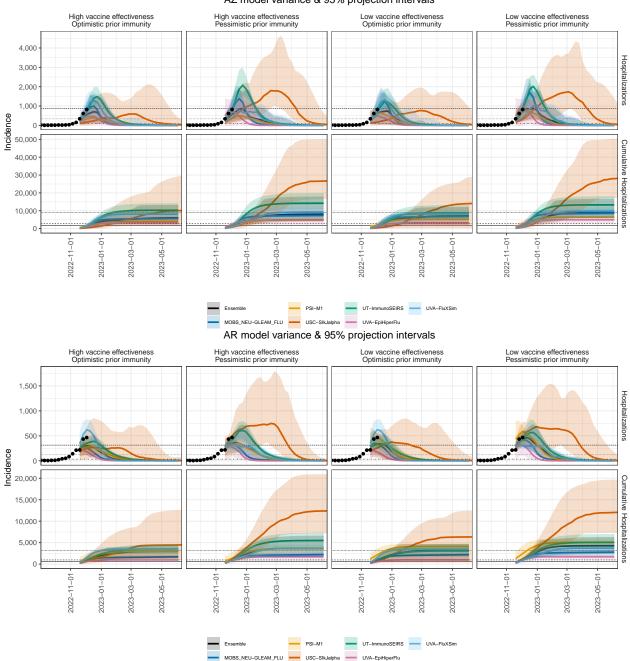




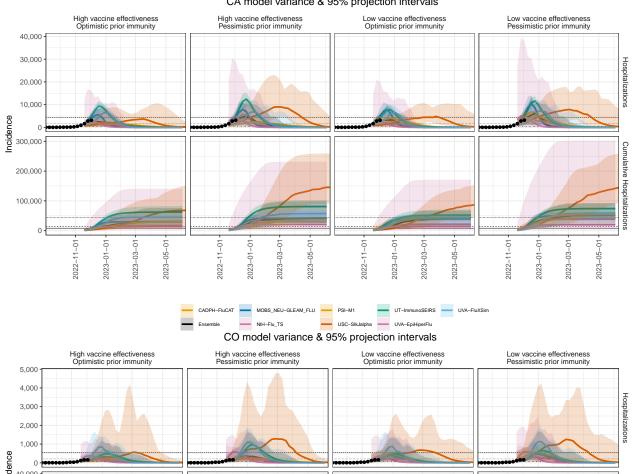
MOBS\_NEU-GLEAM\_FLU

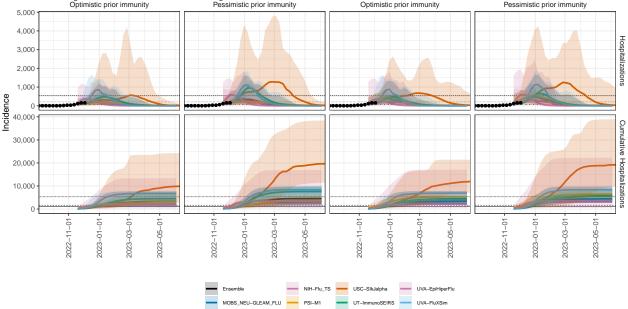
UVA-EpiHiperFlu



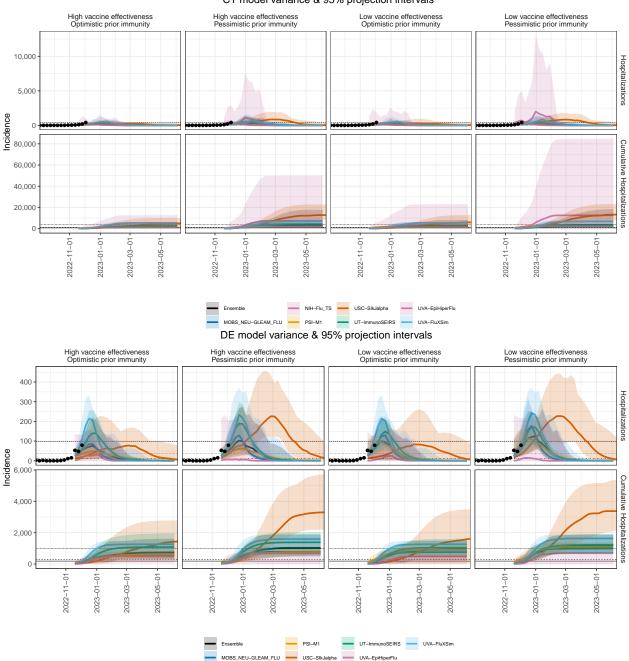


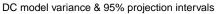


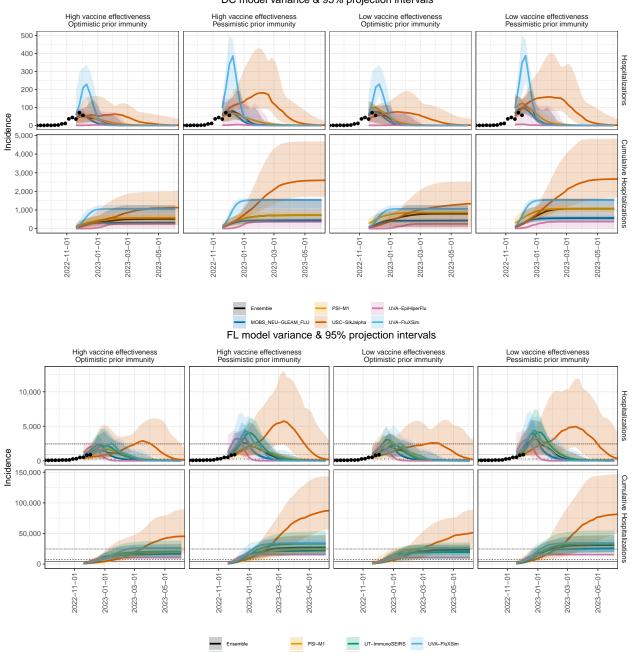






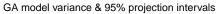


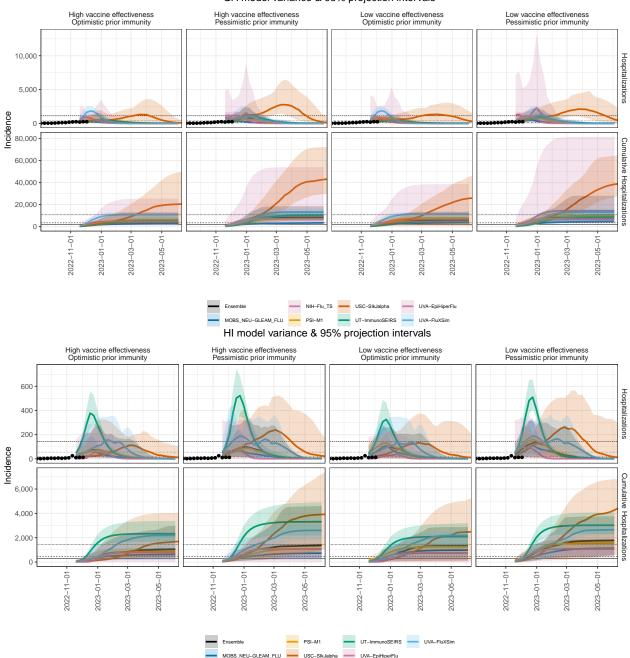




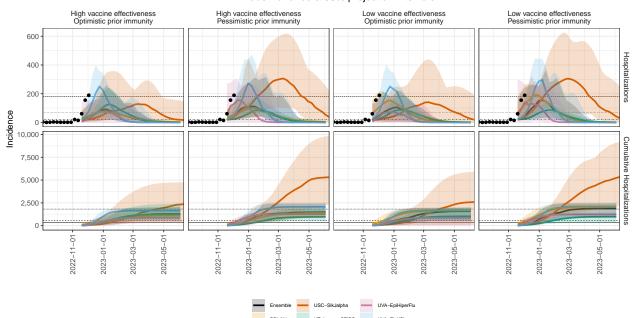
USC-SlkJalpha

UVA-EpiHiperFlu

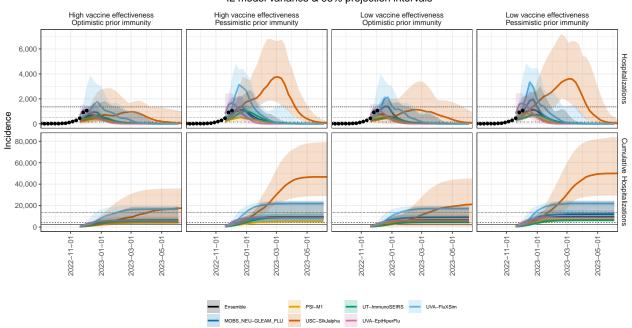




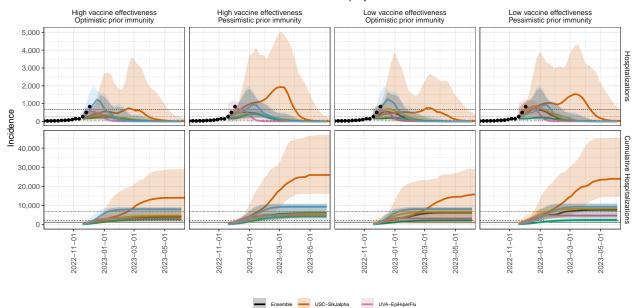
### ID model variance & 95% projection intervals



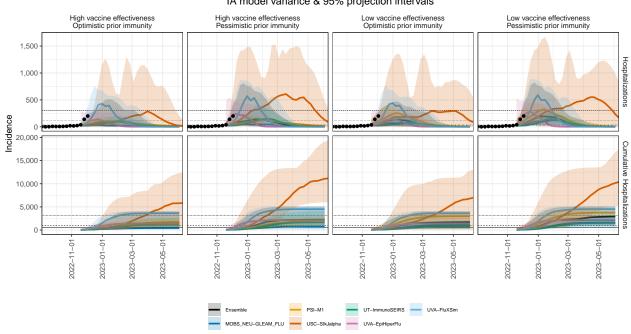
#### IL model variance & 95% projection intervals



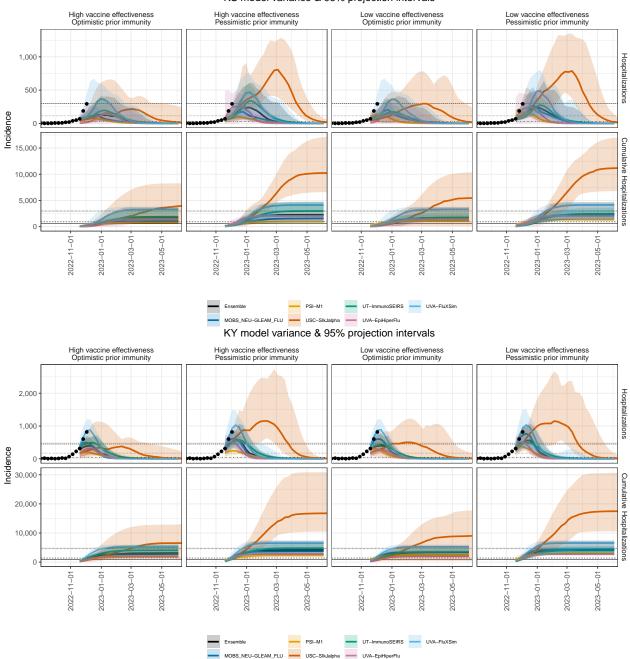
### IN model variance & 95% projection intervals



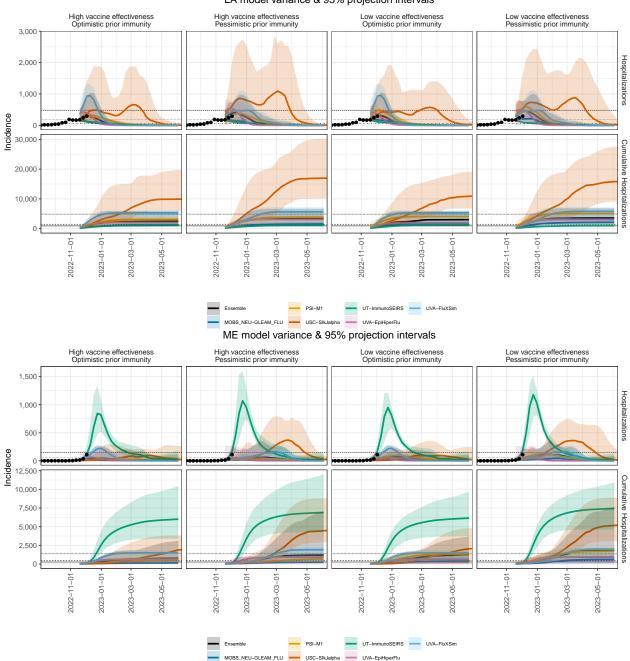
#### IA model variance & 95% projection intervals

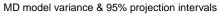


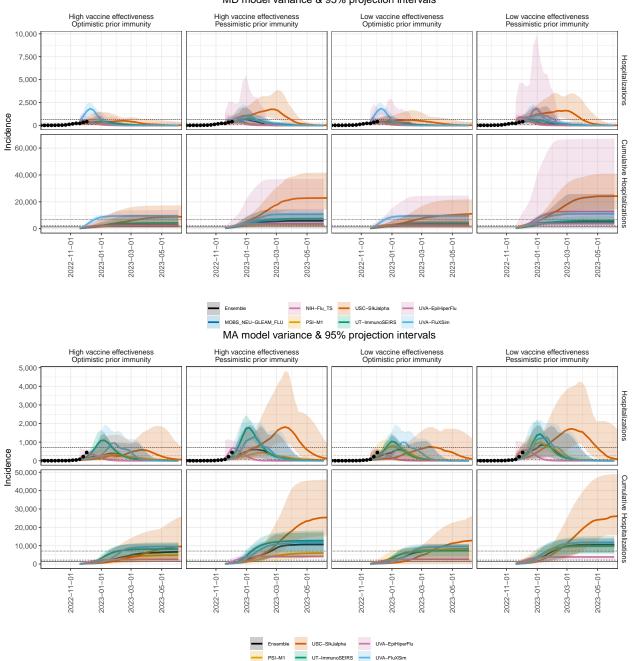
### KS model variance & 95% projection intervals

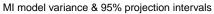


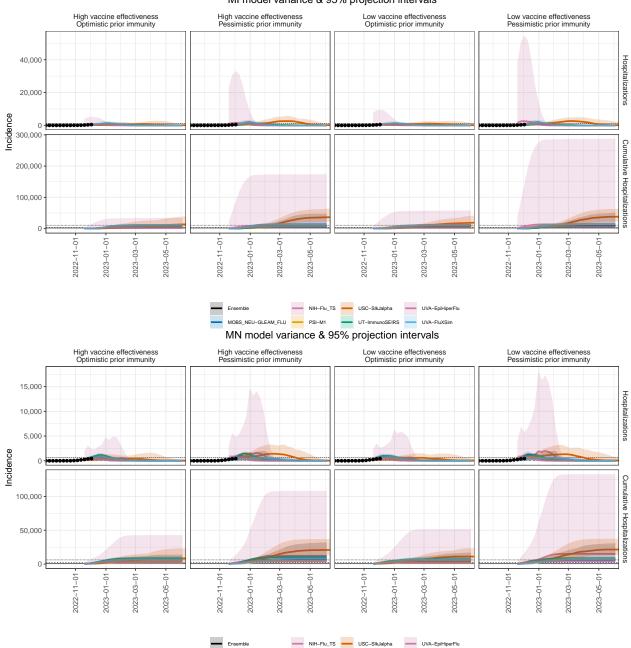


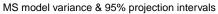


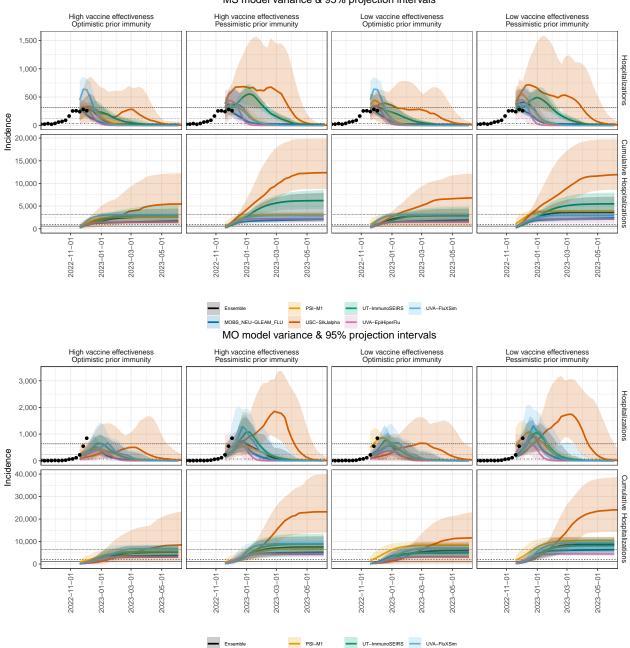




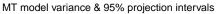


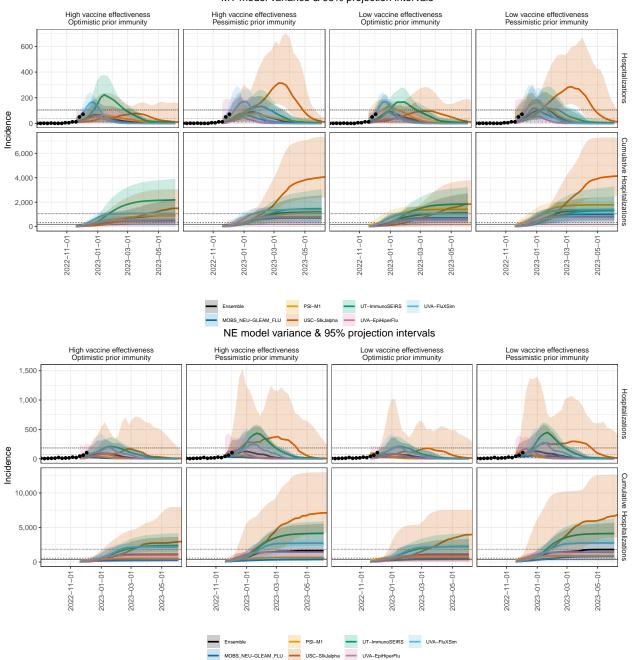


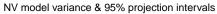


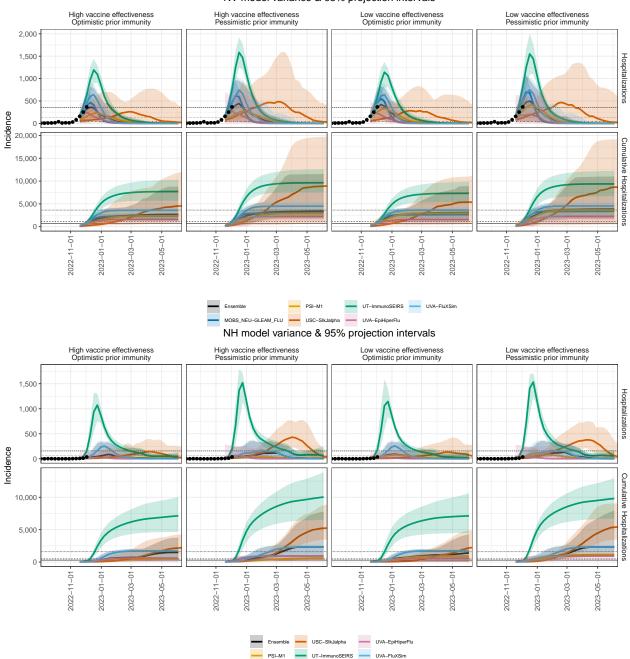


UVA-EpiHiperFlu

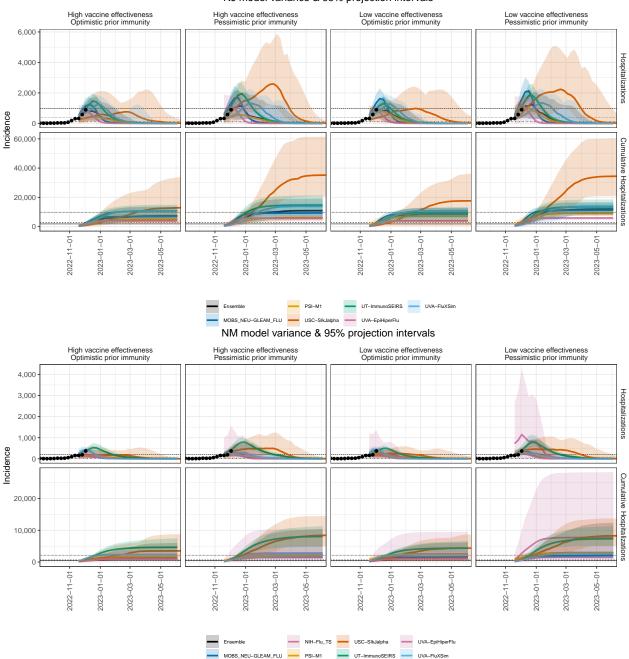




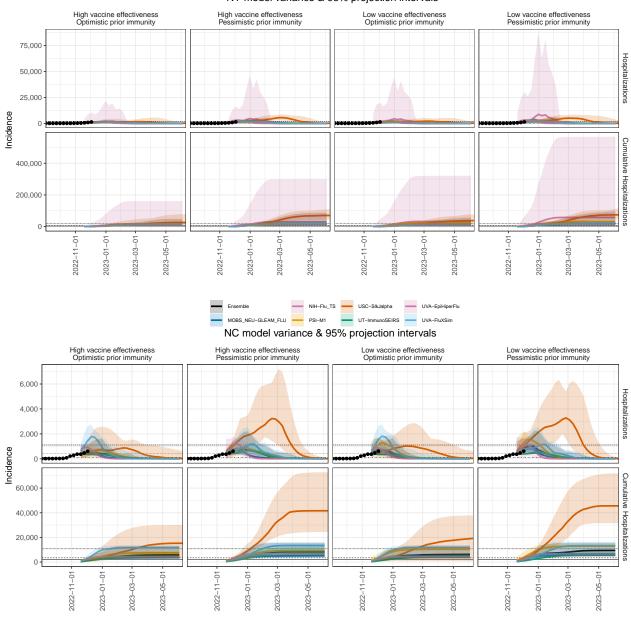








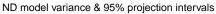


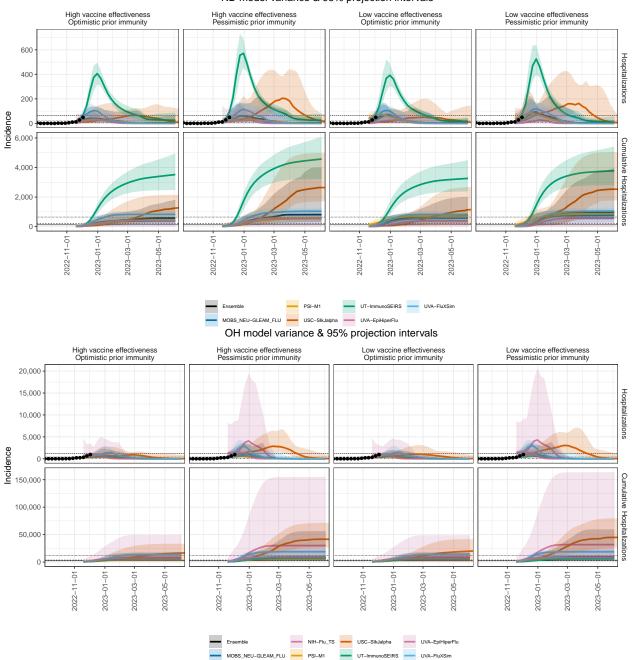


MOBS\_NEU-GLEAM\_FLU

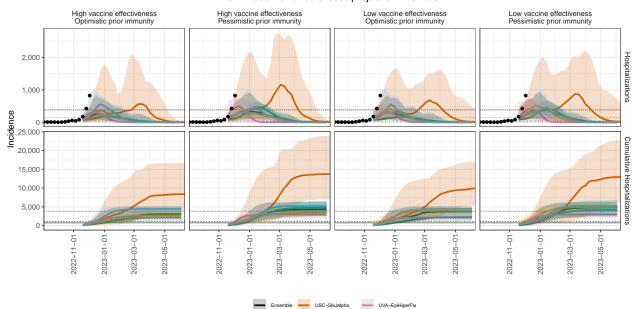
UT-ImmunoSEIRS UVA-FluXSim

UVA-EpiHiperFlu

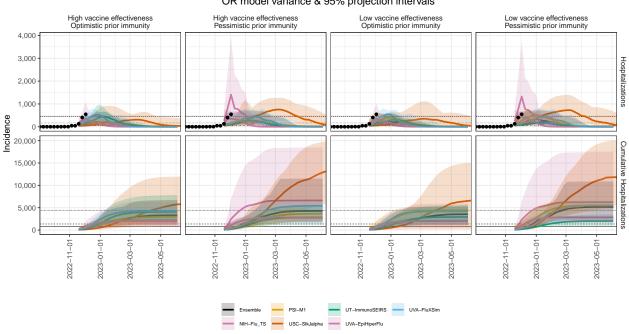




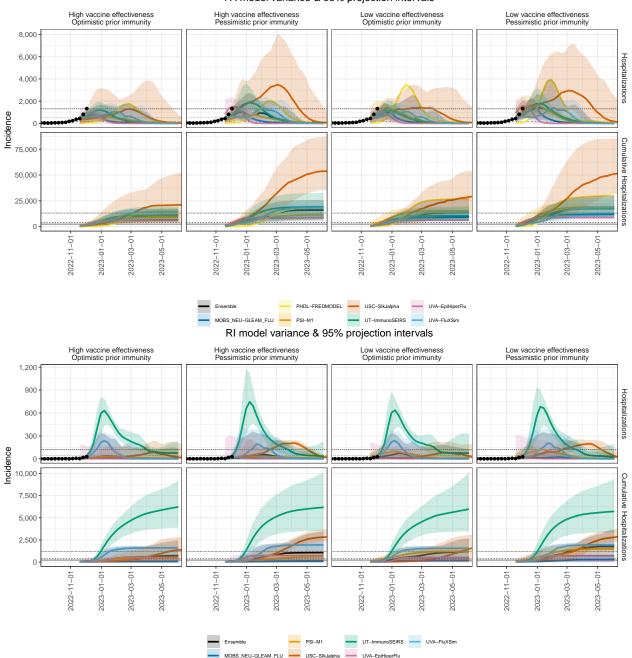
### OK model variance & 95% projection intervals



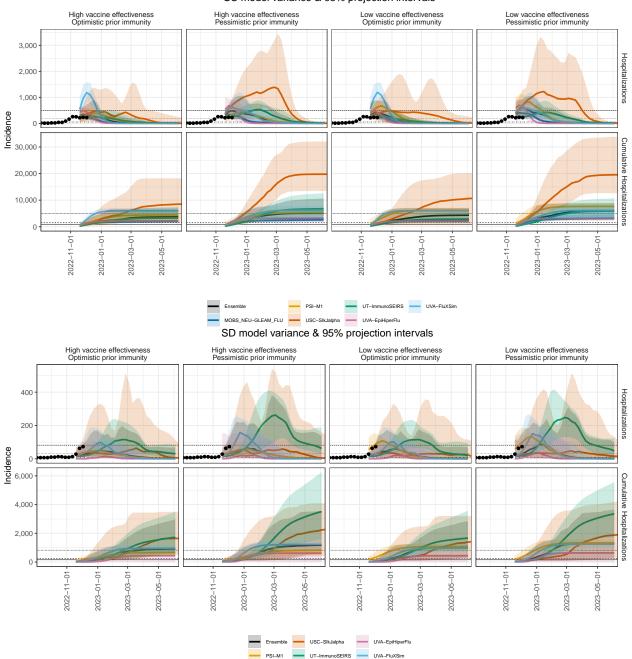
#### OR model variance & 95% projection intervals



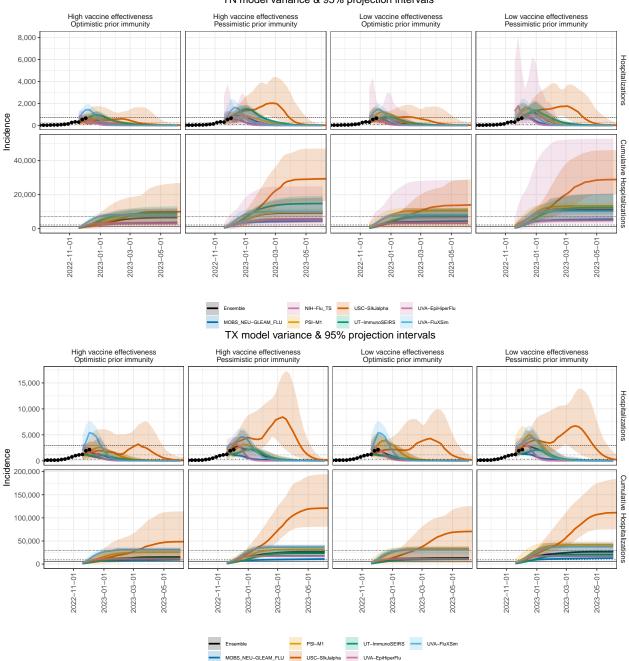
### PA model variance & 95% projection intervals

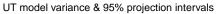


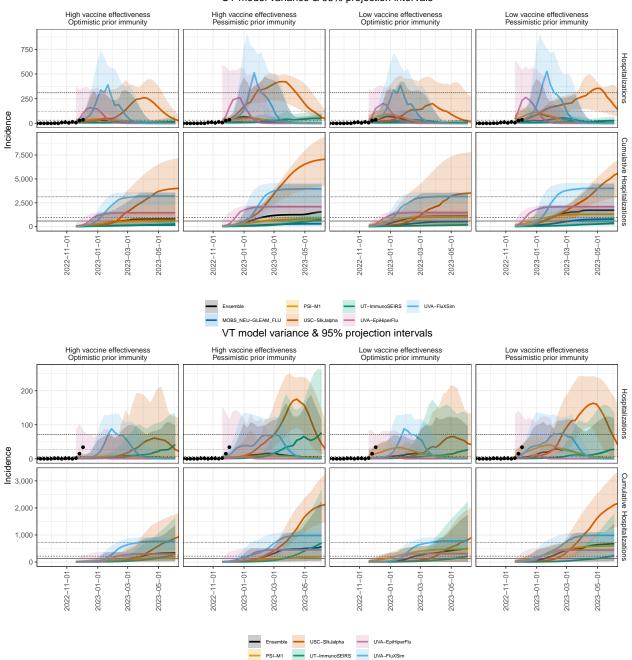
### SC model variance & 95% projection intervals



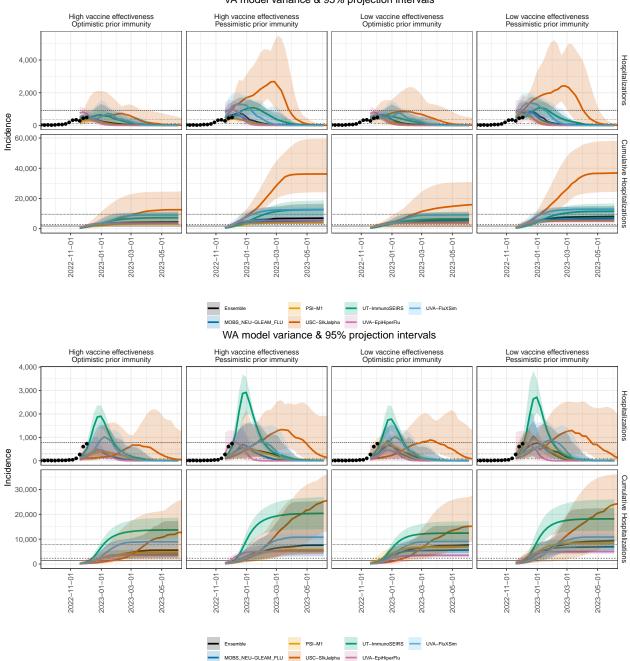
### TN model variance & 95% projection intervals

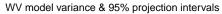


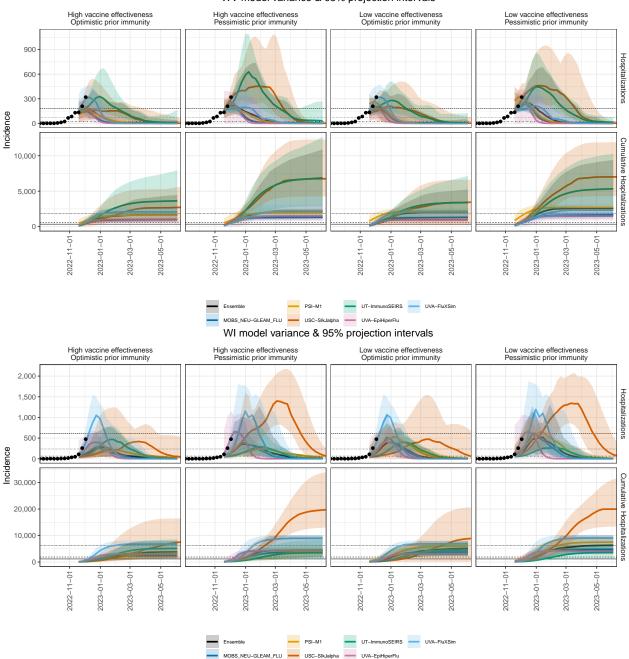


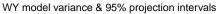


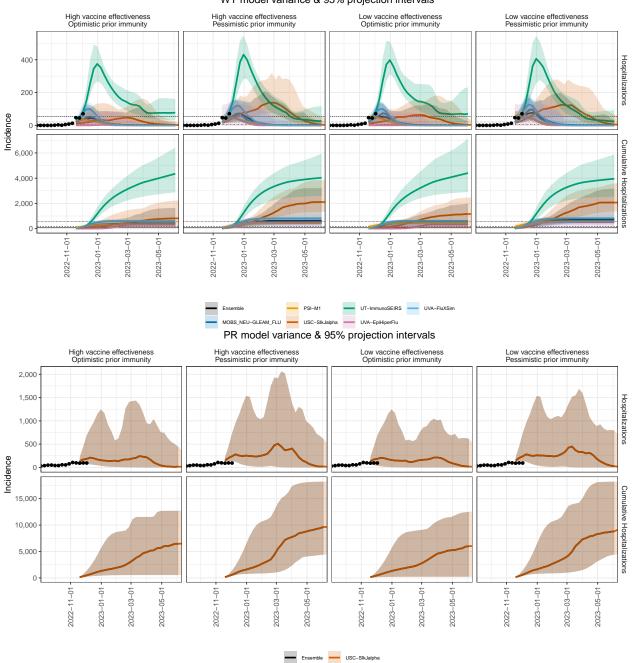












### Teams and models

- California Department of Public Health FluCAT
  - White, L.A. (CADPH), Murray, E. (CADPH), Leon, T.M. (CADPH)
- Center For Disease Dynamics, Economics & Policy FluCompModel
  - Fardad Haghpanah, Eili Klein
- Northeastern University MOBS Lab GLEAM FLU
  - Matteo Chinazzi (Northeastern University, Boston, MA), Jessica T. Davis (Northeastern University, Boston, MA), Kunpeng Mu (Northeastern University, Boston, MA), Alessandro Vespignani (Northeastern University, Boston, MA)
- Fogarty International Center, National Institutes of Health (NIH) Flu TS
  - Amanda Perofsky (NIH), Cécile Viboud (NIH)
- University of Notre Dame FRED
  - Guido Espana, Sean Moore, Alex Perkins
- University of Southern California SIkJalpha
  - Ajitesh Srivastava, Majd Al Aawar
- University of Texas ImmunoSEIRS
  - Kaiming Bi (The University of Texas at Austin), Anass Bouchnita (The University of Texas at El Paso), Spencer J. Fox (The University of Georgia), Lauren Ancel Meyers (The University of Texas at Austin), UT COVID-19 Modeling Consortium.
- University of Virginia Biocomplexity Institute EpiHiper
  - Jiangzhuo Chen (UVA), Stefan Hoops (UVA), Parantapa Bhattacharya (UVA), Dustin Machi (UVA), Bryan Lewis (UVA), Madhav Marathe (UVA)
- University of Virginia Biocomplexity Institute FluXSim
  - Srini Venkatramanan, Aniruddha Adiga, Przemek Porebski, Brian Klahn, Benjamin Hurt, Bryan Lewis (UVA), Madhav Marathe (UVA)
- Fogarty International Center, National Institutes of Health (NIH) FluD
  - Samantha Bents (NIH), Cécile Viboud (NIH)
- Public Health Dynamics Laboratory FREDMODEL
  - Mary G Krauland
- Predictive Science M1
  - Ben-Nun M (Predictive Science), Turtle J (Predictive Science), Riley P (Predictive Science)

### The Flu Scenario Modeling Hub Coordination Team

- Shaun Truelove, Johns Hopkins University
- Cécile Viboud, NIH Fogarty
- Justin Lessler, University of North Carolina
- Sara Loo, Johns Hopkins University
- Lucie Contamin, University of Pittsburgh
- Emily Howerton, Penn State University
- Rebecca Borchering, Penn State University
- Claire Smith, Johns Hopkins University
- Harry Hochheiser, University of Pittsburgh
- Katriona Shea, Penn State University
- Michael Runge, USGS
- Erica Carcelen, Johns Hopkins University
- Sung-mok Jung, University of North Carolina
- J Espino, University of Pittsburgh
- John Levander, University of Pittsburgh