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# Distributed Counterfactual Modeling Approach for Investigating Hospital-Associated Racial Disparities in COVID-19 Mortality

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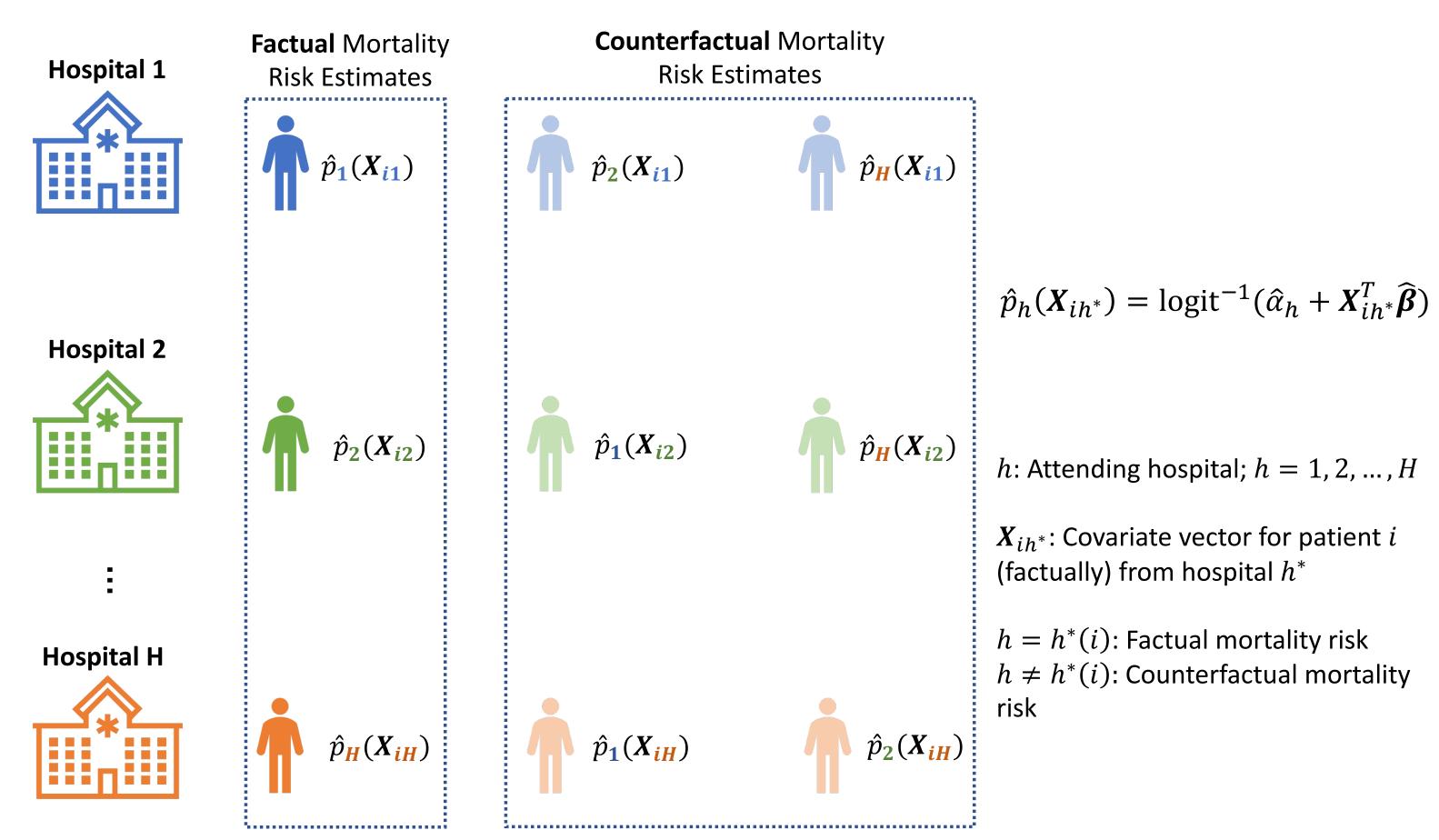
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## Background

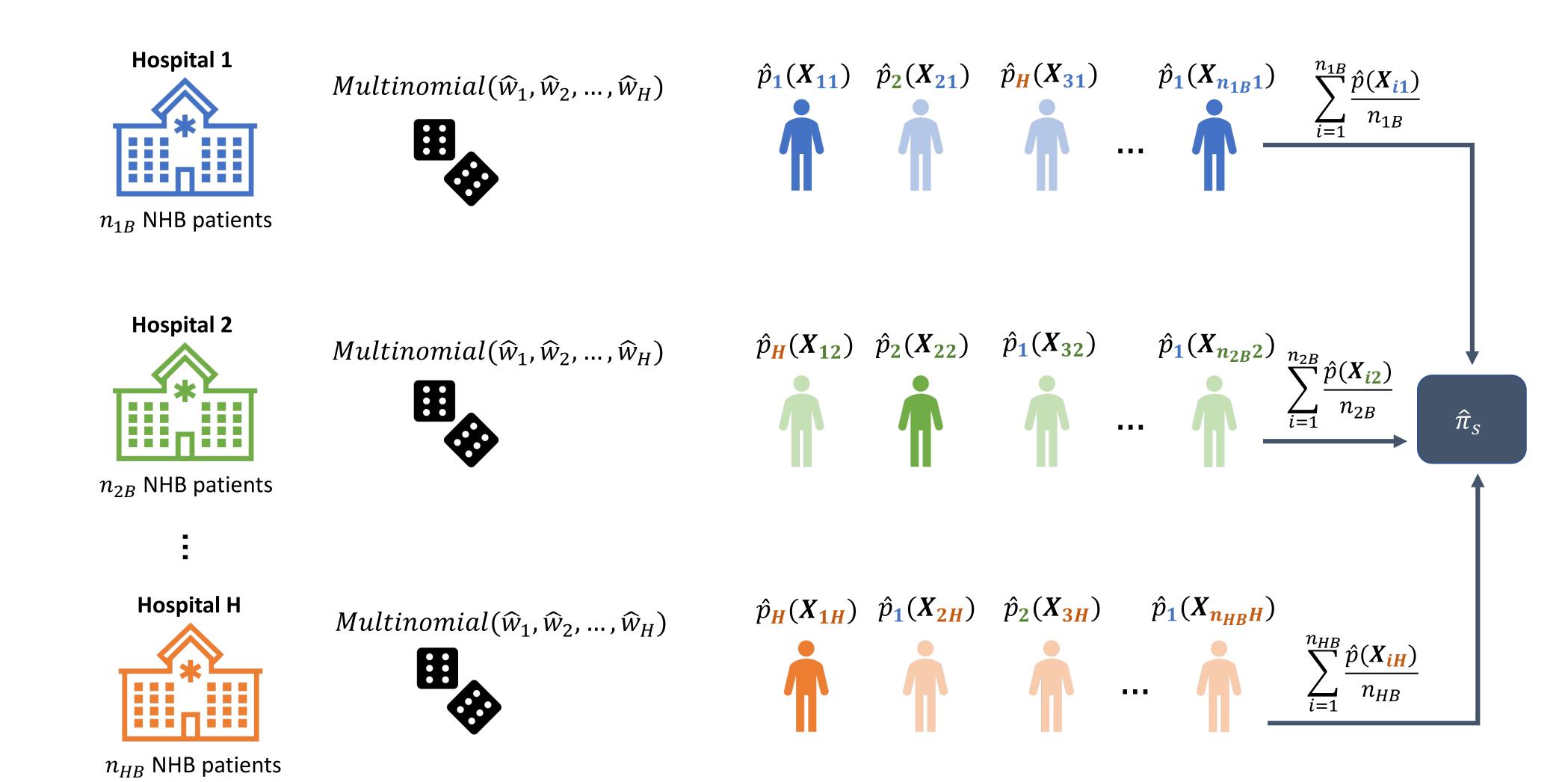
- Several studies have found that black patients are more likely than white patients to test positive for or be hospitalized with COVID-19 but have found no difference in in-hospital mortality.
- Previous studies may have underestimated racial differences due to reliance on data from single hospital system.
- Adequate control of patient-level characteristics requires aggregation of highly granular data from several institutions.
- Want to investigate role of care site in health outcome disparity by race. Patients of different races tend to live in different areas, and sources of care and referral patterns tend to differ.
- Goal: Use counterfactual modeling to study potential association between admitting hospital and racial disparity in mortality for COVID-19 patients without requiring patientlevel data sharing.

### Methods

- Idea: Fit generalized linear mixed model (GLMM) to model log odds of mortality while adjusting for common patient-level fixed effects as well as hospital-specific random effects.
- Counterfactual modeling: Through estimating hospital-specific effects, can estimate patient-specific mortality risk as if patient (counterfactually) attended hospital different from the one truly attended.



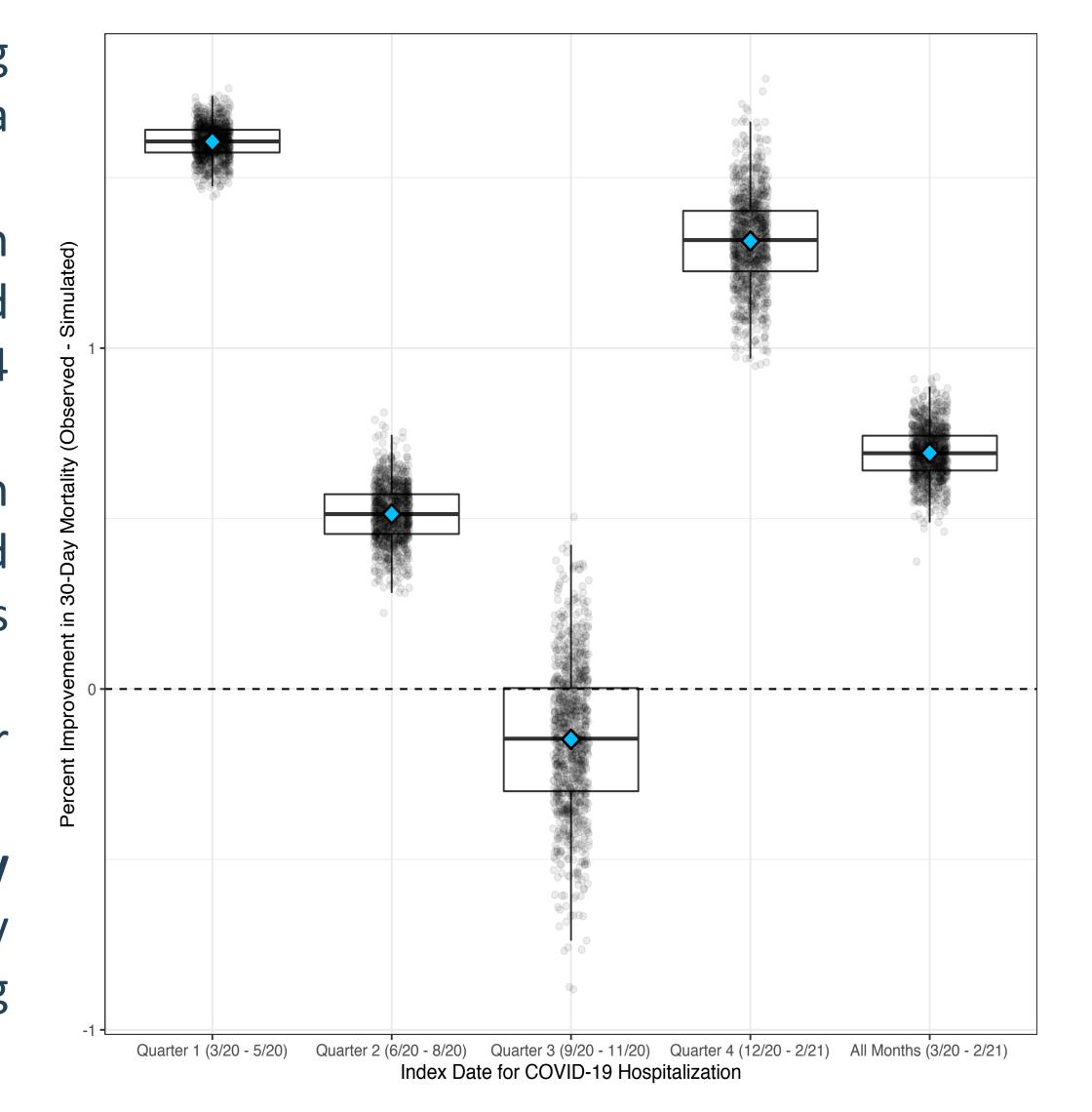
- Distributed penalized quasi-likelihood (dPQL, Luo et al.) algorithm distributively fits GLMM using data stored separately at different hospital systems.
- Aggregate, summary-level information shared rather than patient-level data.
- Simulation used to estimate racial disparity: produce counterfactual mortality rate estimate for black patients had they attended hospitals in the same distribution as white patients (while retaining sociodemographic/clinical characteristics (see schematic overview of simulation procedure on right).



 $\widehat{w}_h$ : proportion of total NHW patients (factually) at hospital h

#### Results

- Proof of concept: Counterfactual modeling simulation using data from OneFlorida Clinical Research Consortium.
- Modeled in-hospital mortality as function of patient characteristics and estimated hospital-specific random effects hospitals).
- difference between Boxplots display observed (factual) mortality rates and average simulated mortality risk estimates for all Non-Hispanic Black patients.
- Performed analysis stratified by quarter and across all quarters.
- Results not meant to be clinically interpreted; intent is to demonstrate utility method performing this counterfactual modeling.



#### Conclusions

- Presented novel application of method for performing distributed generalized linear mixed modeling to study association between admitting hospital and racial disparities in COVID-19 mortality.
- Privacy preserving: requires participating institutions to share only aggregate, summarylevel data to perform counterfactual modeling.
- Potential for this method to be used in OHDSI study to explore racial differences in COVID-19 mortality using data from OHDSI network (as well as in other applications of interest), allowing for more generalizable and clinically impactful conclusions.