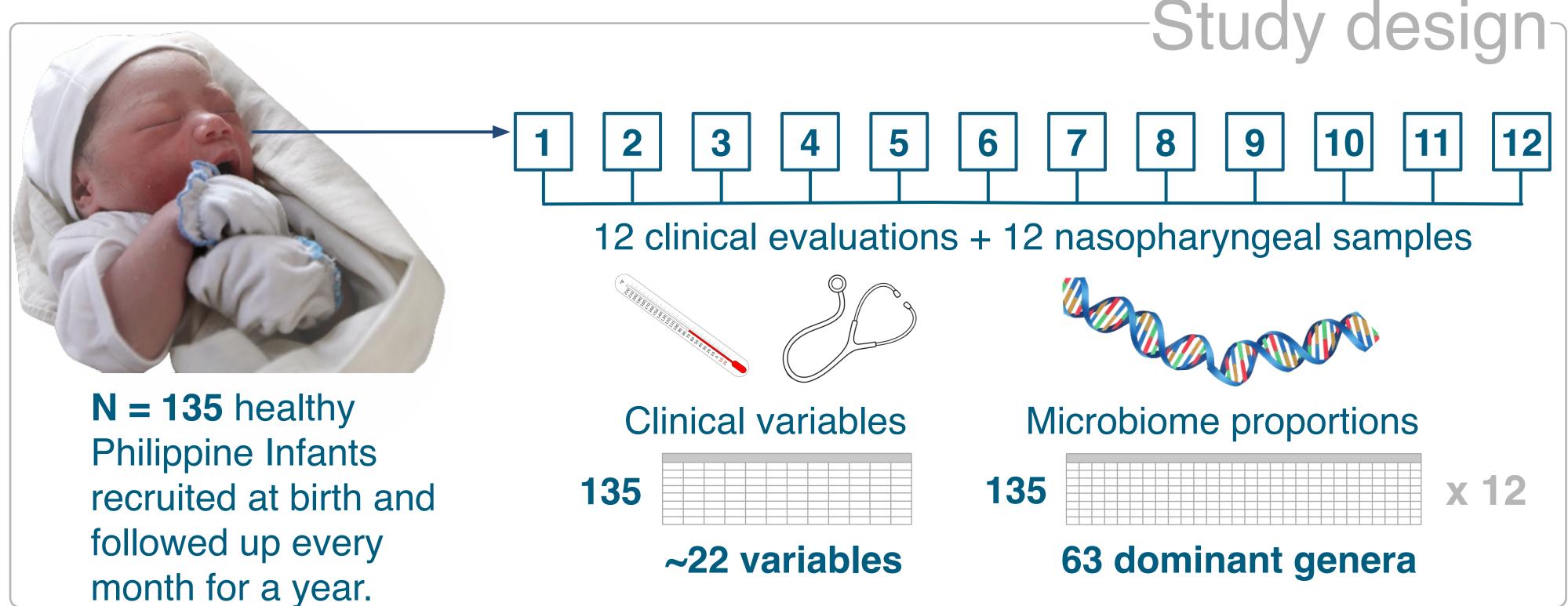
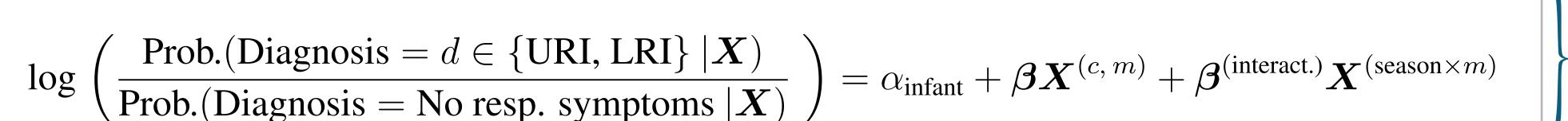
Uncovering microbiome correlates of upper and lower respiratory tract infections in infants using sequential Bayesian analysis

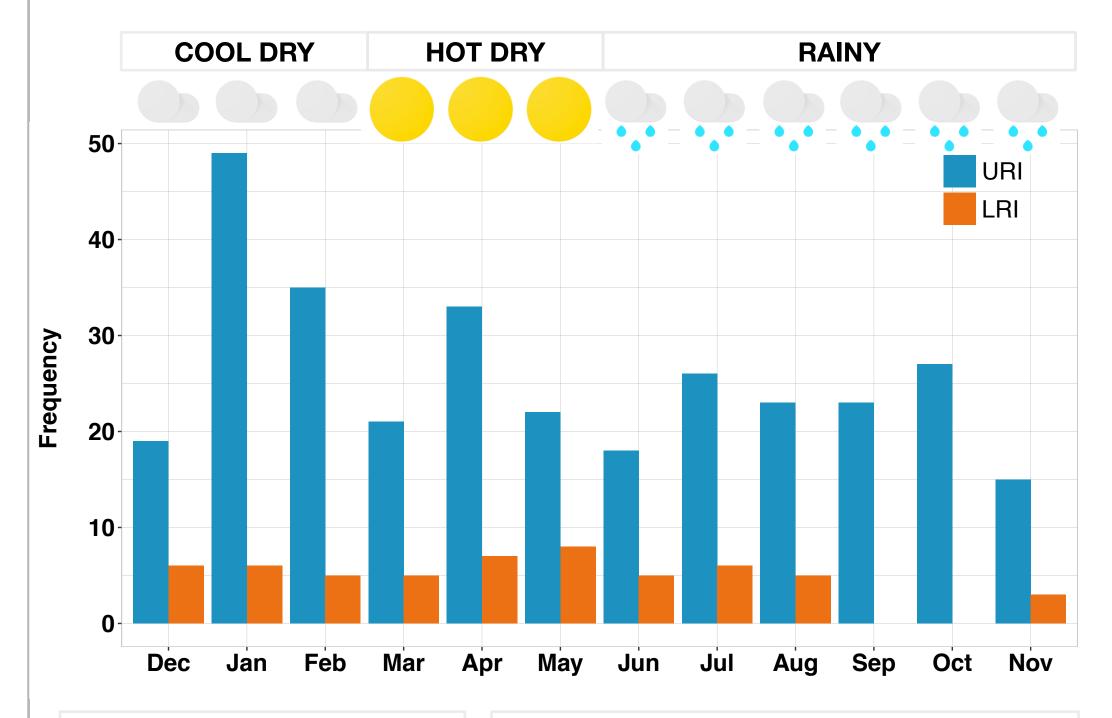
Jyoti Shankar¹, Leilani T. Nillos², Stephanie Mounaud¹, Edelwisa Segubre-Mercado², Aileen M. R. Mojica², Sherry Vi N. Sarra², Xenia D. Geraldino², Herbert Guinto², Vernoni E. Dulalia², Karen E. Nelson¹, Marilla G. Lucero² and William C. Nierman¹. Author affiliations: ¹J. Craig Venter Institute, USA; ²Research Institute for Tropical Medicine, Philippines





Why mixed-effects models with interactions?

Each infant contributes from 4 to 12 samples to the analysis and the distribution of URI and LRI is seasonal in nature.



Model specifications: Multinomial logistic regression with mixed-effects and interaction terms

served as a sanity check.

possible future therapeutic targets.

Sample size: 1474 samples from 135 infants. Response variable: URI, LRI, None (Categorical) Covariates: 63 dominant genera, 22 clinical variables

Mixed-effects model provides a "cross-sectional" view of a time-series dataset while adjusting for repeated measurements from infants.

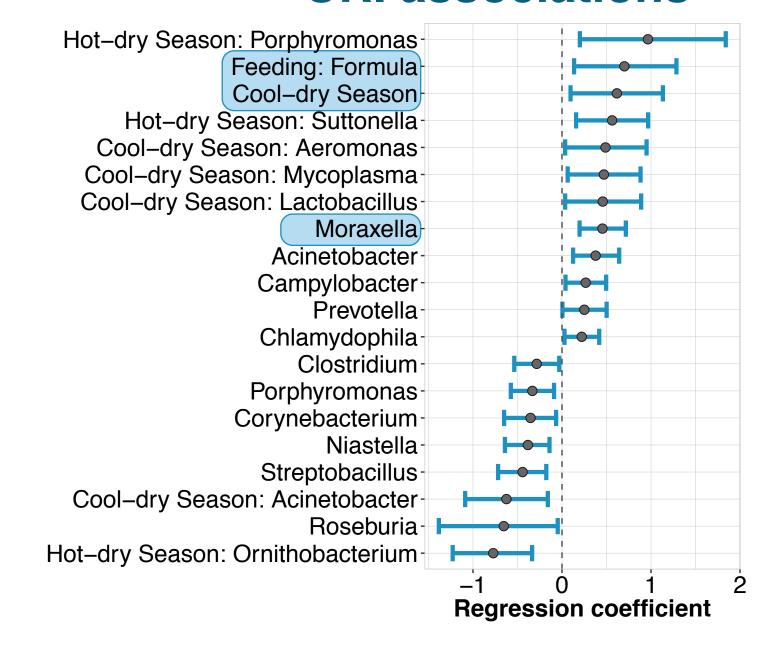
Detected known associations (e.g. known clinical correlates of URI and LRI and known pathogen-containing genera) that

Identified several seasonal-interactions of pathogencontaining as well as commensal microbe groups. Could be mediators of observed seasonal variation in URI/LRI and

Provided estimates of URI and LRI probability for each visit. These estimates were used as the response variable in the time-series models.

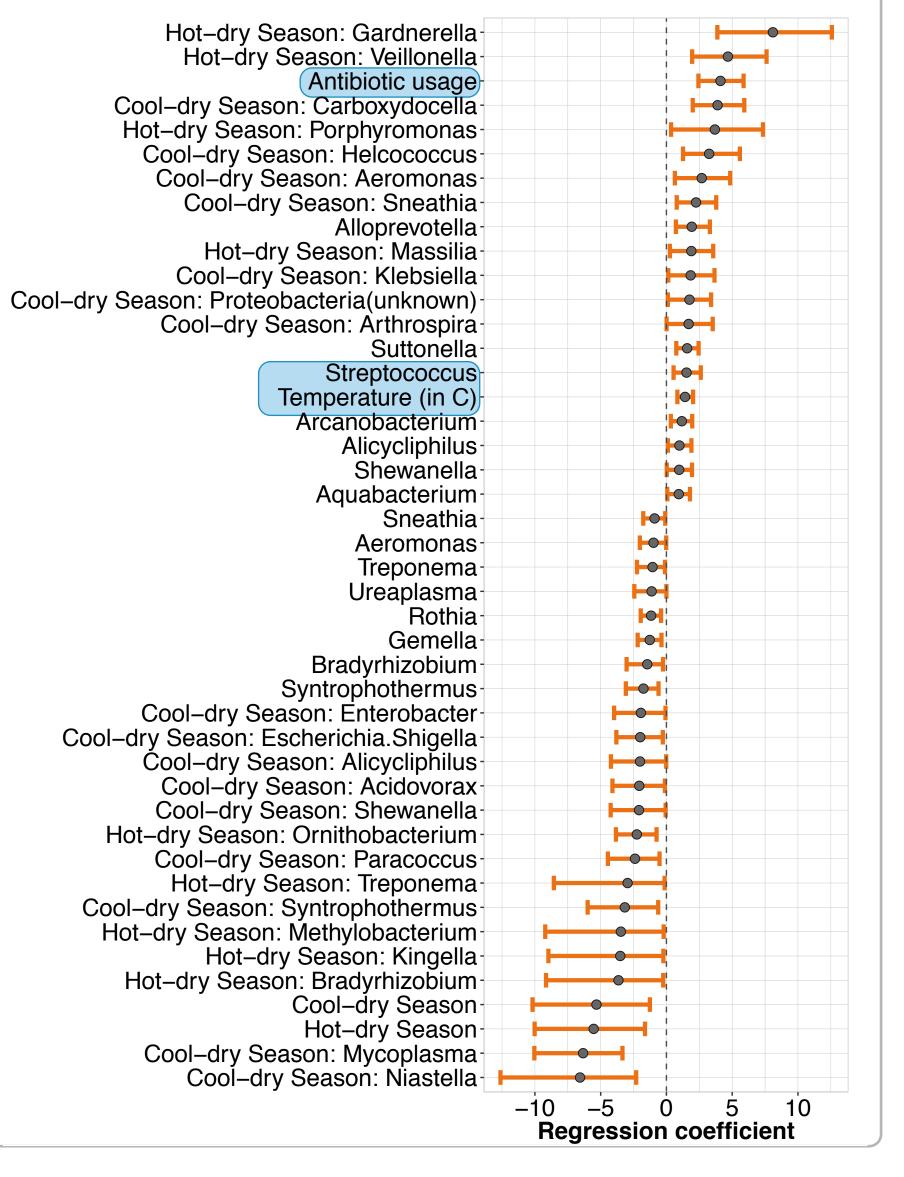
J. Craig Venter® INSTITUTE

URI associations



-Bayesian mixed-effects models

LRI associations



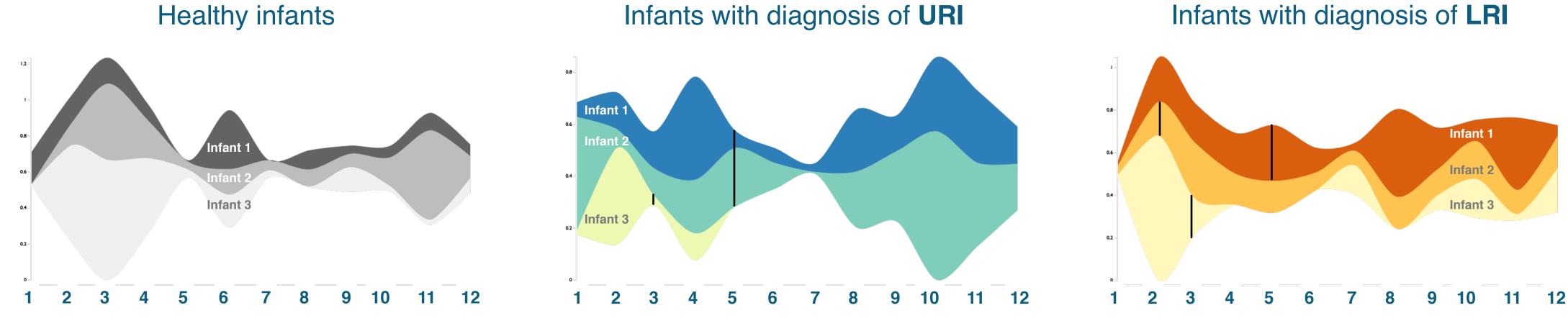
-Background and Objectives

- Upper and lower respiratory tract infections (URI and LRI) are among top illnesses among infants in first year of life.
- The Nasopharynx is the most common pathway of entry of pathogens into the respiratory tract
- To what extent does the microbial community (the microbiome) in the nasopharynx contribute to the establishment and proliferation of pathogens and the occurrence of URI and LRI?

Bayesian structural time-series models

Why infant-level time-series models?

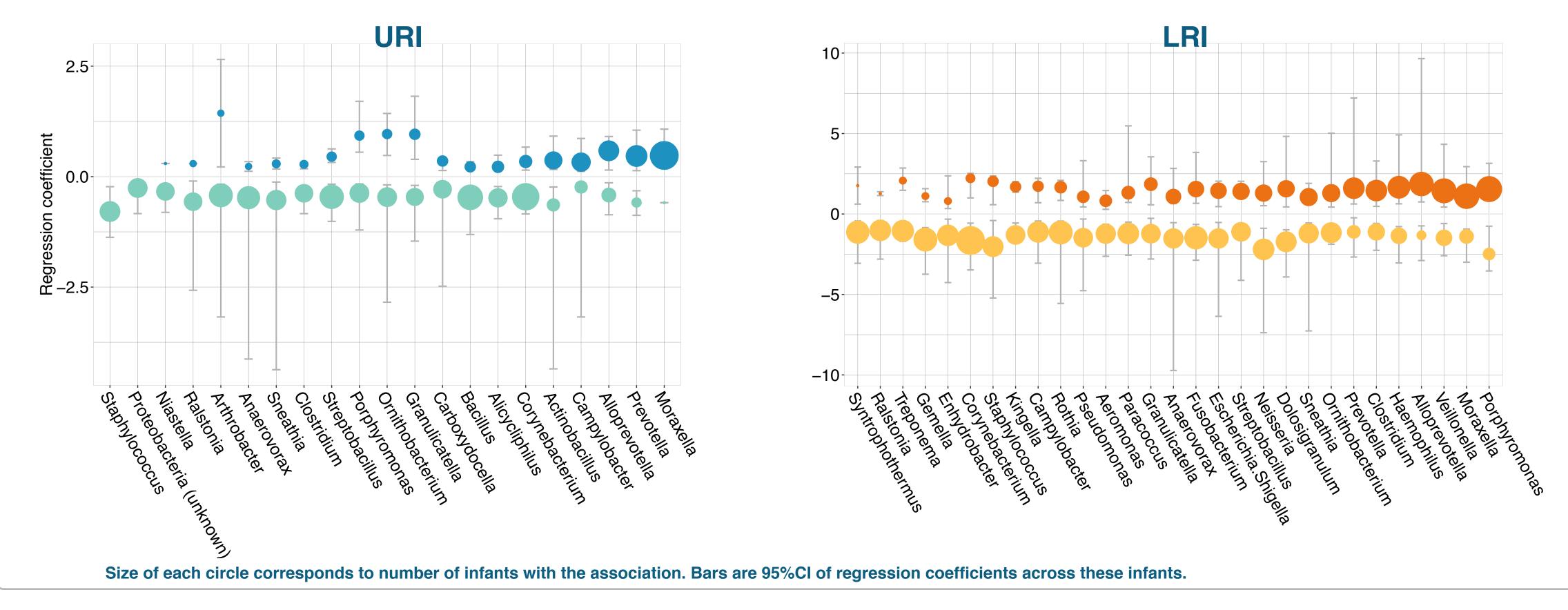
Infant-level heterogeneity in microbe levels is staggering. Below is a tiny example: Streptococcus levels across the 12 visits in 3 representative infants in each group (Healthy, URI, LRI) are shown. Vertical lines represent URI and LRI diagnoses.



Each structural time-series equation modeled the URI/LRI probability across visits as a function of 63 microbial genera and 13 clinical variables (excluding signs and symptoms of URI/LRI). The URI/LRI probability was estimated by the mixed-effects model

$$Y_t = s_t + \beta \boldsymbol{X}_t^{\text{(clinical, microbiome)}} + v_t; v_t \sim \mathcal{N}(0, \sigma_v^2) \dots$$
 (Observation equation) $s_t = s_{t-1} + w_t; w_t \sim \mathcal{N}(0, \sigma_w^2) \dots$ (State equation)

Across 131 infant-specific time-series models, microbial genera significantly associated with either increased or decreased probability of URI and LRI, in at least 20% (≥26) infants are summarized below. URI associations appear to be much more consistent across infants than LRI associations. A handful of prominent pathogen-containing genera and several commensal genera modulate risk of URI. LRI associations are evenly distributed across several genera.



Future directions

- Improve mixed-model estimates of URI/LRI probability and re-estimate the structural time-series models with these improved estimates.
- Adapt survival modeling framework to examine time-to-URI and time-to-LRI events.
- Investigate determinants of recurrent URI and LRI. (Primary challenge: Very sparse data.)

We most gratefully acknowledge funding from the Bill and Melinda Gates Foundation.