# Bayesian Hierarchical Models and Influenza Modeling

Nehemias Ulloa

Iowa State University

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- Approx 4,000,000 cases annually result in severe illness
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- Causes a significant economic and resource burden on the healthcare system
- Vaccines are a simple and effective way of preventing the spread of influenza

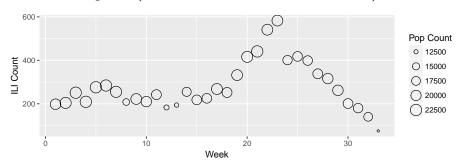
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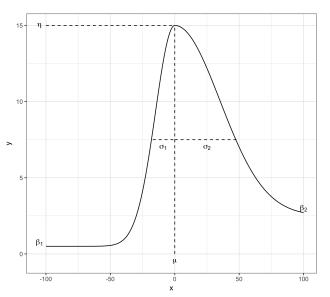


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- Change form of the scaling factor to something borrowed from Werker and Jaggard (1997)

$$ASG(w|\dots) = \begin{cases} \beta_1 + (\eta - \beta_1) \exp[-(w - \mu)^2 / 2\sigma_1^2] & w < \mu \\ \beta_2 + (\eta - \beta_2) \exp[-(w - \mu)^2 / 2\sigma_2^2] & w \ge \mu \end{cases}$$



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$$y_{wrs} \stackrel{ind}{\sim} Bin(n_{wrs}, \phi_{wrs})$$
$$logit(\phi_{wrs}) = ASG(w|\theta_{rs})$$
$$\theta_{rs} \stackrel{ind}{\sim} N(\mu_r, \Delta_r \Omega \Delta_r)$$

$$\mu_r \stackrel{ind}{\sim} N(\mu, \Delta \Omega \Delta) \qquad \mu \stackrel{ind}{\sim} N(m_0, C_0)$$

$$\Delta_r = diag(\sigma_{r,1}, \cdots) \qquad \Delta = diag(\sigma_1, \cdots)$$

$$\sigma_{r,i} \stackrel{ind}{\sim} t_4^+(a, b) \qquad \qquad \sigma_i \stackrel{ind}{\sim} t_4^+(c, d)$$

$$\Omega \stackrel{ind}{\sim} LKJ(n)$$

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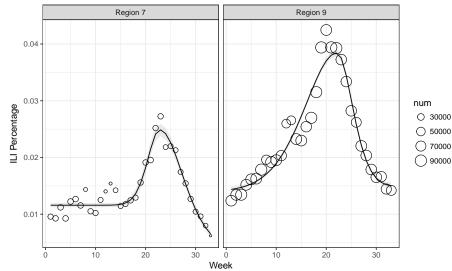
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- Do you need the Asymmetrical Gaussian component?

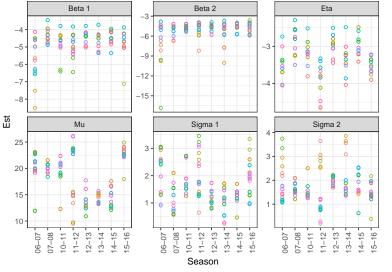
#### Does it work?

### Results for Regions 7 and 9 in the 15-16 Influenza Season:



#### Does it work?

#### What does the form tell us about the Influenza seasons:



#### Region

- Region 1
  - Region 2
- Region 3
- Region 4
- Region 5
- Region 6
- Region 7
- Region 8
- Region 9
- Region
- Region 10

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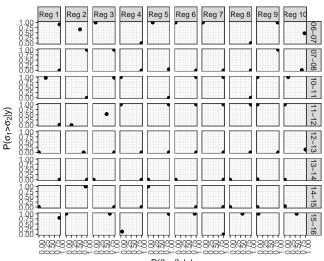
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If  $\beta_1$  and  $\beta_2$ , and  $\sigma_1$  and  $\sigma_2$ , are really not different then the posterior probabilities will hang out around 0.5, but if they hang out around the boundaries, then we need the flexibility.



 $P(\beta_1 > \beta_2 | y)$ 

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- Focus more on forecasting