

Exercises 5 - continued

* How to check hierarchical model - example with cheese problem.

We want to check that the model can produce data similar to the actual data.

For cheese:

$$y_i \sim N(x_i \beta_i, \sigma^2 I)$$

$$\sigma^2 \propto \frac{1}{\sigma^2}$$

$$\beta_i \sim N(\gamma, \sigma^2 I)$$

$$\gamma \propto C$$

$$\Lambda_{ii} \sim IG\left(\frac{1}{2}, \frac{1}{2}\right)$$

From each trace:

→ Take β_i

- Take σ^2
 - Draw y_{ij} for x_i
 - Plot y_0 for each store & compare to real data.
-

A hierarchical probit model

$$P(y_{ij} = 1) = \phi(z_{ij}) \quad \nrightarrow \text{CDF of } N(0, 1)$$

$$z_{ij} = \eta_i + x_{ij}^\top \beta_i$$

$j \equiv$ respondent

$i \equiv$ state

$x_i \Rightarrow$ demographic predictors

$$P(y_{ij} = 1) = P(z_{ij} > 0) = \phi(x_{ij}^\top \beta_i)$$

$$y_{ij} = \mathbb{1}(z_{ij} \geq 0)$$

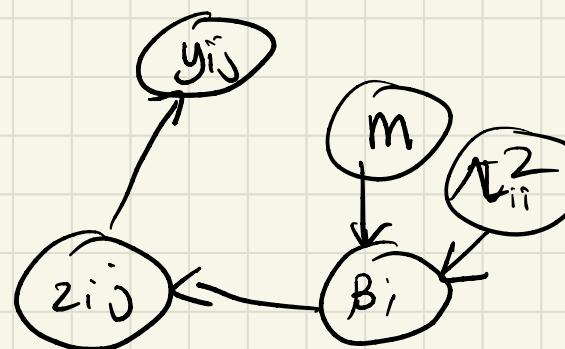
$$z_{ij} \sim N(\underbrace{x_{ij}\beta_i}_\text{includes intercept}, 1)$$

$$\beta_i \sim N(m, \sigma^2 I)$$

$m \propto C$

$$\sigma^2 \propto 16 \left(\frac{1}{2}, \frac{1}{2} \right)$$

* conditionals



$$P(z_{ij} | \cdot) \propto P(y_{ij} | z_{ij}) P(z_{ij})$$

$$= \begin{cases} \mathbb{1}(y_{ij} = 1) N(x_{ij}\beta_i, 1) & \leftarrow \text{truncated normal} \\ \mathbb{1}(y_{ij} = 0) N(x_{ij}\beta_i, 1) \end{cases}$$

$$P(\beta_i | \cdot) \propto P(z_i | \beta_i) P(\beta_i)$$

$$\propto P(\beta_i) \prod_{j=1}^{n_i} P(z_{ij} | \beta_i) P(\beta_i)$$

$$\propto \exp\left(-\frac{1}{2} (\beta_i - m)^T (I^2 I)^{-1} (\beta_i - m)\right)$$

$$\times \exp\left(-\frac{1}{2} (z_i - \beta_i)^T (z_i - \beta_i)\right)$$

$$\propto \exp\left(-\frac{1}{2}\left[(\beta_i - m)^T (\tau^2 I)^{-1} (\beta_i - m)\right.\right.$$

$$\left. + (z_i - x_i^\top \beta_i)^T (z_i - x_i^\top \beta_i) \right)$$

$$\propto \exp\left(-\frac{1}{2}\left(\beta_i^\top (\tau^2 I)^{-1} \beta_i - 2\beta_i^\top (\tau^2 I)^{-1} m + m^\top (\tau^2 I)^{-1}\right.\right.$$

$$m + z_i^\top z_i - 2(\beta_i^\top x_i^\top)(z_i) + (\beta_i^\top x_i^\top)$$

$$\left.\left. - x_i^\top \beta_i\right)\right)$$

$$\propto \exp\left(-\frac{1}{2}\left(\beta_i^\top ((\tau^2 I)^{-1} + x_i^\top x_i) \beta_i\right.\right.$$

$$\left.- 2\beta_i^\top (\tau^2 I)^{-1} m + x_i^\top z_i\right)\left.\right)$$

$$\propto \exp\left(-\frac{1}{2}\left(\beta_i^\top (\hat{m})^\top K^* (\beta_i - m^*)\right)\right)$$

$$K^* = (\tau^2 I)^{-1} + x_i^\top x_i$$

$$m^* = (K^*)^{-1} \left[(\tau^2 I)^{-1} m + x_i^\top z_i \right]$$

$$\beta_i \sim N_p(m^*, (K^*)^{-1})$$

(1x2) (n x 1)

predictions

$$P(m|-) \propto P(m) \prod_{\text{state}} P(\beta_i|m)$$

$$\propto \prod_{\text{state}} \exp \left(-\frac{1}{2} (\beta_i - m)^T (\tau^2 I)^{-1} (\beta_i - m) \right)$$

$$\propto \exp \left(-\frac{1}{2} \sum_{\text{state}} (\beta_i - m)^T (\tau^2 I)^{-1} (\beta_i - m) \right)$$

$$\propto \exp \left(-\frac{1}{2} \sum_{\text{state}} (\beta_i^T (\tau^2 I) \beta_i - 2 m^T (\tau^2 I) m + m^T (\tau^2 I) m) \right)$$

$$+ m^T (\tau^2 I) m$$

$$\sim N_p(m^*, K^*)$$

$$K^* = S (\tau^2 I)^{-1}$$

$$m^* = (K^*)^{-1} \sum_{\text{state}} (\tau^2 I)^{-1} \beta_i$$

$$P(T^2 I) \propto P(T^2) \prod_{\text{state}} P(\beta_i | T^2)$$

$$\propto (T^2)^{-1 - \frac{1}{2}} \exp\left(-\frac{1}{2}(T^2)^{-1}\right) \prod_{\text{state}} \frac{1}{(T^2 + \frac{1}{2})}$$

$$\exp\left(-\frac{1}{2} (\beta_i - m)(T^2 I)^{-1} (\beta_i - m)\right)$$

$$\propto (T^2)^{-1 - \frac{1}{2}} \exp\left(-\frac{1}{2}(T^2)^{-1}\right) \times (T^2)^{-\frac{1}{2} \times S}$$

$$\times \exp\left(-\frac{1}{2} \sum_{\text{state}} (\beta_i - m)(T^2 I)^{-1} (\beta_i - m)\right)$$

$$\propto (T^2)^{-1 - \frac{1}{2}} \exp\left(-\frac{1}{2}(T^2)^{-1}\right)$$

$$\exp\left(-\frac{1}{2T^2} \sum (\beta_i - m)^T (\beta_i - m)\right)$$

$$\propto (T^2)^{-1 - \frac{1}{2} - \frac{S}{2}}$$

$$-\frac{1}{2}(T^2)^{-1} \exp\left(-\frac{1}{2T^2} \sum_{i=1}^S (\beta_i - m)^T (\beta_i - m)\right)$$

$$\propto \left[G\left(\frac{S+1}{2}\right) \right]$$

$$\frac{\sum_{i=1}^S (\beta_i - m)^T (\beta_i - m) + 1}{2}$$

$$\stackrel{(1)}{\sum} (1 \times 12) (12 \times 1)$$