

0 Introduction

The goal of this document is to fully characterize the Dunson day-specific probabilities model. In its current state it tries to provide full detail of the derivations described in *Bayesian Inferences on Predictors of Conception Probabilities*.

1 The model

1.1 Marginal probability of conception

The marginal probability of conception, obtained by integrating out the couple-specific frailty ξ_i , has form as follows.

$$\begin{aligned} \mathbb{P}(Y_{ij} = 1 | \mathbf{X}_{ij}, \mathbf{U}_{ij}) &= \int_0^\infty \mathbb{P}(Y_{ij}, \xi_i | \mathbf{X}_{ij}, \mathbf{U}_{ij}) d\xi_i \\ &= \int_0^\infty \mathbb{P}(Y_{ij}, \xi_i | \mathbf{X}_{ij}, \mathbf{U}_{ij}) \mathcal{G}(\xi_i; \phi, \phi) d\xi_i \\ &= \int_0^\infty \left[1 - \prod_{k=1}^K (1 - \lambda_{ijk})^{X_{ijk}} \right] \mathcal{G}(\xi_i; \phi, \phi) d\xi_i \\ &= 1 - \int_0^\infty \prod_{k=1}^K (1 - \lambda_{ijk})^{X_{ijk}} \mathcal{G}(\xi_i; \phi, \phi) d\xi_i \\ &= 1 - \int_0^\infty \prod_{k=1}^K \left[\exp \left\{ -\xi_i \exp(\mathbf{u}'_{ijk} \boldsymbol{\beta}) \right\} \right]^{X_{ijk}} \mathcal{G}(\xi_i; \phi, \phi) d\xi_i \\ &= 1 - \int_0^\infty \prod_{k=1}^K \exp \left\{ -\xi_i X_{ijk} \exp(\mathbf{u}'_{ijk} \boldsymbol{\beta}) \right\} \mathcal{G}(\xi_i; \phi, \phi) d\xi_i \\ &= 1 - \int_0^\infty \exp \left\{ -\xi_i \sum_{k=1}^K X_{ijk} \exp(\mathbf{u}'_{ijk} \boldsymbol{\beta}) \right\} \mathcal{G}(\xi_i; \phi, \phi) d\xi_i \\ &= 1 - \left[\frac{\phi}{\phi + \sum_{k=1}^K X_{ijk} \exp(\mathbf{u}'_{ijk} \boldsymbol{\beta})} \right]^\phi \end{aligned}$$

since

$$\begin{aligned}
& \int_0^\infty \exp \left\{ -\xi_i \sum_{k=1}^K X_{ijk} \exp(\mathbf{u}'_{ijk} \boldsymbol{\beta}) \right\} \mathcal{G}(\xi_i; \phi, \phi) d\xi_i \\
&= \int_0^\infty \exp \left\{ -\xi_i \sum_{k=1}^K X_{ijk} \exp(\mathbf{u}'_{ijk} \boldsymbol{\beta}) \right\} \frac{\phi^\phi}{\Gamma(\phi)} \xi_i^{\phi-1} d\xi_i \\
&= \int_0^\infty \frac{\phi^\phi}{\Gamma(\phi)} \xi_i^{\phi-1} \exp \left\{ -\xi_i \left[\phi + \sum_{k=1}^K X_{ijk} \exp(\mathbf{u}'_{ijk} \boldsymbol{\beta}) \right] \right\} d\xi_i \\
&= \left[\frac{\phi}{\phi + \sum_{k=1}^K X_{ijk} \exp(\mathbf{u}'_{ijk} \boldsymbol{\beta})} \right]^\phi \int_0^\infty \frac{[\phi + \sum_{k=1}^K X_{ijk} \exp(\mathbf{u}'_{ijk} \boldsymbol{\beta})]^\phi}{\Gamma(\phi)} \\
&\quad \times \xi_i^{\phi-1} \exp \left\{ -\xi_i \left[\phi + \sum_{k=1}^K X_{ijk} \exp(\mathbf{u}'_{ijk} \boldsymbol{\beta}) \right] \right\}^\phi d\xi_i
\end{aligned}$$

2 Posterior computation

Express the data augmentation model as

$$\begin{aligned}
Y_{ij} &= I \left(\sum_{k=1}^K X_{ijk} Z_{ijk} > 0 \right), \\
Z_{ijk} &\sim \text{Poisson} \left(\xi_i \exp(\mathbf{u}'_{ijk} \boldsymbol{\beta}) \right), \quad k = 1, \dots, K
\end{aligned} \tag{1}$$

Let us further define $W_{ijk} = X_{ijk} Z_{ijk}$ for all i, j, k .

2.1 Verifying the equivalence of the data augmentation model