

We introduce the models for zero inflated count data. Define the ZINB model for the response variable y assuming only non-negative values as follows:

$$f_{ZINB}(y) = \phi I(y = 0) + (1 - \phi) f_D(y),$$

where ϕ denotes the mixture proportion and $I(\cdot)$ is the indicator variable defined as

$$I(y) = \begin{cases} 1, & \text{if } y = 0 \\ 0, & \text{if } y \sim f_D(\cdot) \end{cases},$$

and $f_D(\cdot)$ is the density for Negative Binomial/Poisson distribution given by

$$f_{NB}(y) = \frac{\Gamma(y + 1/\alpha)}{y! \Gamma(1/\alpha)} \left(\frac{\alpha^{-1}}{\alpha^{-1} + \lambda} \right)^{\alpha^{-1}} \left(\frac{\lambda}{\alpha^{-1} + \lambda} \right)^y$$

$$f_{Pois}(y) = \frac{e^{-\lambda} \lambda^y}{y!}$$

where $\lambda (> 0)$ is the mean of y and $Var(y) = \lambda$ for Poisson and $\lambda + a\lambda^2$ for negative Binomial, $a \geq 0$ being the dispersion parameter for negative Binomial distribution. It is easy to see that the mean and variance of y under the ZINB model is given by

$$E(y) = (1 - \phi) \lambda, \quad Var(y) = [(1 - \phi) \lambda] [1 + \lambda(\phi + \alpha)]$$

while those for the ZIP model is

$$E(y) = (1 - \phi) \lambda, \quad Var(y) = [(1 - \phi) \lambda] [1 + \lambda\phi]$$

The mean parameter λ is related to the covariates $\mathbf{x}_i = (x_0, x_1, \dots, x_p)_i$ by the log link function given by

$$\log(\lambda_i) = \sum_{j=0}^p x_j \beta_j = \mathbf{x}_i' \boldsymbol{\beta}$$

where $\boldsymbol{\beta} = (\beta_0, \beta_1, \dots, \beta_p)$ is the $p + 1$ dimensional vector of unknown regression coefficients. Finally the mixture proportion ϕ is linked to the set of zero inflated covariates $\mathbf{z} = (z_0, z_1, \dots, z_q)$ and the $q + 1$ vector of regression coefficients $\boldsymbol{\gamma} = (\gamma_0, \gamma_1, \dots, \gamma_q)$ by the logit link function as in the following

$$\text{logit}(\phi) = \sum_{k=0}^q z_k \gamma_k = \mathbf{z}' \boldsymbol{\gamma}.$$

The z_0 is assumed to be 1 corresponding to the intercept γ_0 .