

## A Appendix

Let  $f(y_i; \mu_i, \phi)$  be the density function for a negative binomial distribution with mean  $\mu_i$  and dispersion parameter  $\phi$ :

$$f(y_i; \mu_i, \phi) = \frac{\Gamma(y_i + 1/\phi)}{y_i! \Gamma(1/\phi)} \left( \frac{1}{1 + \phi \mu_i} \right)^{1/\phi} \left( \frac{\phi \mu_i}{1 + \phi \mu_i} \right)^{y_i}. \quad (9)$$

As  $\phi \rightarrow 0$ ,  $f(y_i; \mu_i, \phi)$  converges to a Poisson distribution with mean  $\mu_i$ . The log likelihood is

$$l(\mathbf{y}, \boldsymbol{\mu}, \phi) = \sum_{i=1}^n \left[ \log \left( \frac{\Gamma(y_i + 1/\phi)}{y_i! \Gamma(1/\phi)} \right) - \left( \frac{1}{\phi} + y_i \right) \log(1 + \phi \mu_i) + y_i \log(\phi \mu_i) \right].$$

where  $\mathbf{y} = (x_1, \dots, x_n)^\top$ ,  $\boldsymbol{\mu} = (\mu_1, \dots, \mu_n)^\top$ , and  $n$  indicates sample size. We further assume

$$\log(\mu_i) = b_0 + \sum_{j=1}^J x_{ij} b_j$$

and maximize the penalized log likelihood

$$l(\mathbf{y}, \boldsymbol{\mu}, \phi) - \sum_{j=1}^J (1 + \delta) \log(|b_j| + \tau).$$

The Iteratively Reweighted Least Squares (IRLS) approach is often used to maximize the log likelihood of a generalized linear model by solving a series of least squares problem. Given the current parameter estimates  $\tilde{b}_j$  ( $0 \leq j \leq J$ ) and  $\tilde{\phi}$ , the negative (unpenalized) log likelihood can be written as a quadratic form

$$l_Q(\mathbf{y}, \boldsymbol{\mu}, \phi) = - \sum_{i=1}^n w_i \left[ z_i - b_0 - \sum_{j=1}^J x_{ij} b_j \right]^2,$$

where  $z_i = \tilde{\mu}_i + (y_i - \tilde{\mu}_i)/v(y_i)$ ,  $\tilde{\mu}_i = \tilde{b}_0 + \sum_{j=1}^J x_{ij} \tilde{b}_j$ ,  $v(y_i) = \tilde{\mu}_i + \tilde{\mu}_i^2 \tilde{\phi}$ , and  $w_i = v(y_i)$ . Next, given new estimates of  $b_j$  ( $0 \leq j \leq J$ ),  $\phi$  can be re-estimated by a maximize conditional likelihood of  $\phi$ .

Similar to the approach proposed by Friedman et al. [Friedman et al., 2010], We employed an ECM algorithm [Sun et al., 2010] named iterative adaptive Lasso (IAL) to solve this problem of penalized generalized linear model, which include four levels of loops:

- Loop 1 (outer loop): Iterate all combinations of tuning parameter  $\delta$  and  $\tau$ .
- Loop 2: Alternately updates regression coefficients  $b_j$  ( $0 \leq j \leq J$ ) and over dispersion parameter  $\phi$ .
- Loop 3: Update the quadratic approximation  $l_Q$  using current estimate of  $b_j$  ( $0 \leq j \leq J$ ) and  $\phi$ .
- Loop 4 (inner loop): Run the IAL to solve the penalized weighted least squares problem.