

# Some Formulas

GLIM

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$$f(y | \theta_i, \phi) = \exp\left\{ [y\theta_i - b(\theta_i)] / \left(\frac{\phi}{w_i}\right) + c_i(y, \phi) \right\}$$

$$(a) b'(\theta_i) = E(Y_i) \equiv \mu_i$$

$$(b) \left(\frac{\phi}{w_i}\right) b''(\theta_i) = \text{var}(Y_i)$$

$$g(\mu_i) \equiv \eta_i = \begin{cases} x_i^T \beta \\ x_i^T \beta + c_i \leftarrow \text{offset (Poisson)} \end{cases}$$

canonical link:  $\eta_i = \theta_i$

$$L = \log\text{-lik} = \frac{1}{\phi} \sum_{i=1}^n w_i [y_i \theta_i - b(\theta_i)] + \sum_{i=1}^n c(y_i, \phi)$$

$\Rightarrow$

$$\frac{\partial L}{\partial \beta_j} = \frac{1}{\phi} \sum_{i=1}^n w_i [y_i - b'(\theta_i)] \cdot \frac{\partial \theta_i}{\partial \beta_j}$$

$$\frac{\partial^2 L}{\partial \beta_k \partial \beta_j} = -\frac{1}{\phi} \sum_{i=1}^n w_i \left\{ b''(\theta_i) \left(\frac{\partial \theta_i}{\partial \beta_k}\right) \left(\frac{\partial \theta_i}{\partial \beta_j}\right) - [y_i - b'(\theta_i)] \frac{\partial^2 \theta_i}{\partial \beta_k \partial \beta_j} \right\}$$

Using  $\frac{\partial \theta_i}{\partial \beta_j} = \frac{d\theta_i}{d\mu_i} \cdot \frac{d\mu_i}{d\eta_i} \cdot \frac{\partial \eta_i}{\partial \beta_j} = \frac{1}{b''(\theta_i)} \cdot \frac{1}{g'(\mu_i)} \cdot x_{ij}$  with (a), (b):

$$\frac{\partial L}{\partial \beta_j} = \sum_{i=1}^n (y_i - \mu_i) \cdot \frac{1}{g'(\mu_i) \cdot \text{var}(Y_i)} \cdot x_{ij}$$

$$\frac{\partial^2 L}{\partial \beta_k \partial \beta_j} = -\sum_{i=1}^n \frac{x_{ij} x_{ik}}{\text{var}(Y_i) [g'(\mu_i)]^2} + \underbrace{\frac{1}{\phi} \sum_{i=1}^n w_i (y_i - \mu_i) \frac{\partial^2 \theta_i}{\partial \beta_k \partial \beta_j}}_{II}$$

Fisher scoring:  $\hat{\beta}_{(m+1)} = \hat{\beta}_{(m)} + (X^T A_m^{-1} X)^{-1} X^T A_m^{-1} \zeta_m$  where

$$A = \text{diag}[g'(\mu_i)^2 \cdot \text{var}(Y_i)], \quad \zeta = (g'(\mu_i) \cdot (y_i - \mu_i))$$

Asymptotic Distribution:

$$(\hat{\beta} - \beta) \approx M \cdot (\underline{y} - \underline{\mu}) \text{ for } M = (X^T A^{-1} X)^{-1} X^T A^{-1} G, \quad G = \text{diag}(g'(\mu_i))$$

$$\Rightarrow \hat{\beta} \approx N(\beta, W) \text{ for } W = \left\{ X^T [\text{diag}(g'(\mu_i)^2 \cdot \text{var}(Y_i))]^{-1} X \right\}^{-1}$$