

# The score function

To obtain the maximum likelihood estimator for the parameter  $\beta_j$  we derive the score function using the chain rule:

$$\frac{d\ell}{d\beta_j} = U_j = \sum_{i=1}^N \left[ \frac{d\ell_i}{d\beta_j} \right] = \sum_{i=1}^N \left[ \frac{d\ell_i}{d\theta_i} \cdot \frac{d\theta_i}{d\mu_i} \cdot \frac{d\mu_i}{d\beta_j} \right]. \quad (3.1.3)$$

Now, let's consider each term separately

- $\frac{d\ell_i}{d\theta_i} = Y_i b'(\theta_i) + c'(\theta_i) = b'(\theta_i)(Y_i - \mu_i)$
- $\frac{d\theta_i}{d\mu_i} = \left( \frac{d\mu_i}{d\theta_i} \right)^{-1} = \left( \frac{-c''(\theta_i)}{b'(\theta_i)} + \frac{c'(\theta_i)b''(\theta_i)}{[b'(\theta_i)]^2} \right)^{-1} = (b'(\theta_i) \text{Var}(Y_i))^{-1}$
- $\frac{d\mu_i}{d\beta_j} = \frac{d\mu_i}{d\eta_i} \cdot \frac{d\eta_i}{d\beta_j} = \frac{d\mu_i}{d\eta_i} X_{ij}$

Hence, the score function is

$$U_j = \sum_{i=1}^N \left[ \frac{(y_i - \mu_i)}{\text{Var}(Y_i)} X_{ij} \left( \frac{d\mu_i}{d\eta_i} \right) \right] \quad (3.1.4)$$

