

## Log-likelihood equations for LEXPIT model

Under the LEXPIT, the model for the binomial probability is

$$\pi = x' \beta + \text{expit}(z' \gamma)$$

for  $x$  and  $z$  covariates, with  $z$  including an intercept term and ‘expit’ denotes the sigmoid (inverse-logit) function  $\text{expit}(x) = \exp(x)/(1 + \exp(x))$ .

For  $n$  observations and the  $i$ th event outcome as  $y_i$  and case weight  $w_i$ , the score for an arbitrary coefficient  $\theta$  is

$$S(\theta) = \sum_i \dot{\pi}(\theta; x_i, z_i) w_i \nu(\pi(\theta; x_i, z_i))^{-1} (y_i - \pi(\theta; x_i, z_i)),$$

where  $\nu(x) = x(1-x)$  the first derivative of the expit function. Let  $A_i(\theta) = \nu(\pi(\theta; x_i, z_i))^{-1} (y_i - \pi(\theta; x_i, z_i))$ . The Hessian is

$$\dot{S}(\theta) = \sum_i w_i [\ddot{\pi}(\theta; x_i, z_i) A_i(\theta) - \dot{\pi}(\theta; x_i, z_i) \dot{\pi}(\theta; x_i, z_i)' \nu(\pi(\theta; x_i, z_i))^{-1} (A_i(\theta) + 1)].$$

For  $\pi(\beta; x_i, z_i)$ ,

$$\dot{\pi}(\beta; x_i, z_i) = x_i \text{ and } \ddot{\pi}(\beta; x_i, z_i) = 0.$$

For  $\pi(\gamma; x_i, z_i)$ ,

$$\dot{\pi}(\gamma; x_i, z_i) = z_i \text{dexpit}(z_i' \gamma)$$

and

$$\ddot{\pi}(\gamma; x_i, z_i) = z_i z_i' \text{ddexpit}(z_i' \gamma),$$

where  $\text{dexpit}(x) = \text{expit}(x)(1 - \text{expit}(x))$  and  $\text{ddexpit}(x) = \text{dexpit}(x)(1 - 2\text{expit}(x))$ . The Hessian for  $\beta$  is

$$\dot{S}(\beta) = - \sum_i w_i x_i x_i' \nu(\pi(\theta; x_i, z_i))^{-1} (A_i(\theta) + 1).$$

and for  $\gamma$  is

$$\dot{S}(\gamma) = \sum_i w_i z_i z_i' [\text{ddexpit}(z_i' \gamma) A_i(\theta) - \text{dexpit}(z_i' \gamma)^2 \nu(\pi(\theta; x_i, z_i))^{-1} (A_i(\theta) + 1)],$$

and the partial derivative of  $S(\beta)$  with respect to  $\gamma$  is

$$\frac{\partial S(\beta)}{\partial \gamma} = - \sum_i w_i x_i z_i' [\text{dexpit}(z_i' \gamma) \nu(\pi(\theta; x_i, z_i))^{-1} (A_i(\theta) + 1)].$$

In matrix form,

$$\mathcal{H}(\beta) = -X'W(\beta)X$$

with  $W(\beta) = \text{Diag}\{w_1\nu(\pi(\theta; x_1, z_1))^{-1}(A_1(\theta) + 1), \dots\}$ .

$$\mathcal{H}(\gamma) = Z'W(\gamma)Z$$

with  $W(\gamma) = \text{Diag}\{w_1\text{ddexpit}(z_1'\gamma)A_1(\theta) - \text{dexpit}(z_1'\gamma)^2\nu(\pi(\theta; x_1, z_1))^{-1}(A_1(\theta) + 1), \dots\}$ .

$$\mathcal{H}(\beta, \gamma) = -X'W(\beta, \gamma)Z$$

with  $W(\beta, \gamma) = \text{Diag}\{w_1\text{dexpit}(z_1'\gamma)\nu(\pi(\theta; x_1, z_1))^{-1}(A_1(\theta) + 1), \dots\}$ .