

### Definition of ordinal regression

The dependent random variable  $Y$  has ordinal level of measurement. The number of possible values that  $Y$  can take is  $m + 1$ . A realization of  $Y$  is denoted by  $y$  and  $y \in \{0, 1, \dots, m\}$ . The conditional probability distribution of  $Y$  given a set of interval predictors  $\mathbf{x} = (x_1, \dots, x_k)'$  is  $P(Y = y | \mathbf{x})$ . The regression of  $Y$  on  $\mathbf{x}$  is ordinal if and only if

$$P(Y \leq y | \mathbf{x}) = \sum_{t=0}^y P(Y = t | \mathbf{x}), \quad \text{for all } y, \quad (1)$$

is a monotonic function of  $\mathbf{x}$ . If higher values of  $y$  are associated with higher values of  $\mathbf{x}$ , then this function should be non-increasing, for all  $y$ . If  $P(Y \leq y | \mathbf{x})$  is a monotonic function of  $x_i$ , then its derivative with respect to  $x_i$  is either non-negative or non-positive for all  $x_i$ .

### Multinomial logistic regression

Let  $z_s = 1$  if  $y = s$  and  $z_s = 0$  otherwise, for  $s = 1, \dots, m$ , then the conditional probability distribution of  $Y$  given  $\mathbf{x}$  can be written as

$$P(Y = y | \mathbf{x}) = P(Y = 0 | \mathbf{x}) \prod_{s=1}^m \left\{ \frac{P(Y = s | \mathbf{x})}{P(Y = 0 | \mathbf{x})} \right\}^{z_s}, \quad (2)$$

where

$$P(Y = 0 | \mathbf{x}) = \left\{ 1 + \sum_{s=1}^m \frac{P(Y = s | \mathbf{x})}{P(Y = 0 | \mathbf{x})} \right\}^{-1}. \quad (3)$$

In the multinomial logistic regression model,

$$\frac{P(Y = s | \mathbf{x})}{P(Y = 0 | \mathbf{x})} = \exp(\alpha_{0s} + \boldsymbol{\alpha}'_s \mathbf{x}), \quad \text{for } s = 1, \dots, m, \quad (4)$$

where  $\boldsymbol{\alpha}_s = (\alpha_{1s}, \dots, \alpha_{ks})'$ , so that

$$P(Y = y | \mathbf{x}) = \frac{\exp \left\{ \sum_{s=1}^m z_s (\alpha_{0s} + \boldsymbol{\alpha}'_s \mathbf{x}) \right\}}{1 + \sum_{s=1}^m \exp(\alpha_{0s} + \boldsymbol{\alpha}'_s \mathbf{x})}. \quad (5)$$

It follows that

$$P(Y \leq y | \mathbf{x}) = \begin{cases} \frac{1}{1 + \sum_{s=1}^m \exp(\alpha_{0s} + \boldsymbol{\alpha}'_s \mathbf{x})}, & \text{for } y = 0, \\ \frac{1 + \sum_{s=1}^y \exp(\alpha_{0s} + \boldsymbol{\alpha}'_s \mathbf{x})}{1 + \sum_{s=1}^m \exp(\alpha_{0s} + \boldsymbol{\alpha}'_s \mathbf{x})}, & \text{for } y > 0. \end{cases} \quad (6)$$

The multinomial logistic regression model has  $(k + 1)m$  parameters.

### Adjacent categories model

In general, we have

$$\frac{P(Y = s | \mathbf{x})}{P(Y = 0 | \mathbf{x})} = \prod_{u=1}^s \frac{P(Y = u | \mathbf{x})}{P(Y = u - 1 | \mathbf{x})}, \text{ for } s = 1, \dots, m. \quad (7)$$

In an adjacent categories model,

$$\frac{P(Y = u | \mathbf{x})}{P(Y = u - 1 | \mathbf{x})} = \exp(\beta_{0u} + \boldsymbol{\beta}'_u \mathbf{x}), \quad (8)$$

where  $\boldsymbol{\beta}_u = (\beta_{1u}, \dots, \beta_{ku})'$ , so that  $\alpha_{0s} = \sum_{u=1}^s \beta_{0u}$  and  $\boldsymbol{\alpha}_s = \sum_{u=1}^s \boldsymbol{\beta}_u$ . This general adjacent categories model is just a reparameterization of the multinomial logistic regression model. Under the adjacent categories model, it follows that

$$\frac{P(Y = u | \mathbf{x})}{P(Y = u - 1 | \mathbf{x}) + P(Y = u | \mathbf{x})} = \frac{\exp(\beta_{0u} + \boldsymbol{\beta}'_u \mathbf{x})}{1 + \exp(\beta_{0u} + \boldsymbol{\beta}'_u \mathbf{x})}. \quad (9)$$

### Ordinal logistic regression

#### *Adjacent categories*

In the ordinal logistic adjacent categories model,  $\beta_{is} = \beta_i$ , for  $i = 1, \dots, k$  and all  $s$ , so that  $\alpha_{is} = s\beta_i$  and

$$P(Y = y | x_1, \dots, x_k) = \frac{\exp\left(\sum_{s=1}^m z_s \alpha_{0s} + y \boldsymbol{\beta}' \mathbf{x}\right)}{1 + \sum_{s=1}^m \exp(\alpha_{0s} + s \boldsymbol{\beta}' \mathbf{x})}, \quad (10)$$

where  $\boldsymbol{\beta} = (\beta_1, \dots, \beta_k)'$ . The ordinal logistic adjacent categories model has  $k + m$  parameters.

*Cumulative probabilities*

In general, we have

$$P(Y = y | \mathbf{x}) = \begin{cases} P(Y \leq 0 | \mathbf{x}), & \text{for } y = 0, \\ P(Y \leq y | \mathbf{x}) - P(Y \leq y - 1 | \mathbf{x}), & \text{for } y = 1, \dots, m - 1, \\ 1 - P(Y \leq m - 1 | \mathbf{x}), & \text{for } y = m. \end{cases}$$

In the ordinal logistic cumulative probabilities model,

$$P(Y \leq y | \mathbf{x}) = \frac{1}{1 + \exp\{\alpha_{0y} + \boldsymbol{\beta}'\mathbf{x}\}}, \quad \text{for } y = 0, 1, \dots, m - 1, \quad (11)$$

where  $\boldsymbol{\beta} = (\beta_1, \dots, \beta_k)' > 0$ . This model has also  $k + m$  parameters.