

# Econometrics Homework 2

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## Question 1

We have the following ordinal regression model:

$$\begin{aligned} z_i &= x_i' \beta + \epsilon_i \quad \forall i = 1, \dots, n \\ \gamma_{j-1} < z_i \leq \gamma_j &\implies y_i = j, \quad \forall i, j = 1, \dots, J \end{aligned}$$

where (in the first equation)  $z_i$  is the latent variable for individual  $i$ ,  $x_i$  is a vector of covariates,  $\beta$  is a  $k \times 1$  vector of unknown parameters, and  $n$  denotes the number of observations. The second equation shows how  $z_i$  is related to the observed discrete response  $y_i$ , where  $-\infty = \gamma_0 < \gamma_1 < \gamma_{J-1} < \gamma_J = \infty$  are the cut-points (or thresholds) and  $y_i$  is assumed to have  $J$  categories or outcomes.

(a)

We assume that  $\epsilon_i \sim N(0, 1)$ , for  $i = 1, 2, \dots, n$ . Therefore we have, The probability of success,

$$\begin{aligned} Pr(y_i = j) &= Pr(\gamma_{j-1} < z_i \leq \gamma_j) \\ &= Pr(\gamma_{j-1} < x_i' \beta + \epsilon_i \leq \gamma_j) \\ &= Pr(\gamma_{j-1} - x_i' \beta < \epsilon_i \leq \gamma_j - x_i' \beta) \\ &= \Phi(\gamma_j - x_i' \beta) - \Phi(\gamma_{j-1} - x_i' \beta) \quad [\text{where } \Phi(\cdot) \text{ is the cdf of } N(0, 1)] \end{aligned}$$

The likelihood function for the ordinal probit model is,

$$\begin{aligned} L(\beta, \gamma; y) &= \prod_{i=1}^n \prod_{j=1}^J Pr(y_i = j | \beta, \gamma)^{I(y_i=j)} \quad [\text{where } I(\cdot) \text{ is the indicator function}] \\ &= \prod_{i=1}^n \prod_{j=1}^J (\Phi(\gamma_j - x_i' \beta) - \Phi(\gamma_{j-1} - x_i' \beta))^{I(y_i=j)} \end{aligned}$$

(b)

We assume that  $\epsilon_i \sim \mathcal{L}(0, 1)$ , for  $i = 1, 2, \dots, n$ . Therefore the probability of success,

$$\begin{aligned} Pr(y_i = j) &= Pr(\gamma_{j-1} < z_i \leq \gamma_j) \\ &= Pr(\gamma_{j-1} < x_i' \beta + \epsilon_i \leq \gamma_j) \\ &= Pr(\gamma_{j-1} - x_i' \beta < \epsilon_i \leq \gamma_j - x_i' \beta) \\ &= \frac{1}{1 + e^{-(\gamma_j - x_i' \beta)}} - \frac{1}{1 + e^{-(\gamma_{j-1} - x_i' \beta)}} \quad [\text{as } \epsilon_i \sim \mathcal{L}(0, 1)] \\ &= \frac{e^{-(\gamma_{j-1} - x_i' \beta)} - e^{-(\gamma_j - x_i' \beta)}}{(1 + e^{-(\gamma_j - x_i' \beta)})(1 + e^{-(\gamma_{j-1} - x_i' \beta)})} \\ &= \frac{e^{x_i' \beta} (e^{-\gamma_{j-1}} - e^{-\gamma_j})}{(1 + e^{-(\gamma_j - x_i' \beta)})(1 + e^{-(\gamma_{j-1} - x_i' \beta)})} \end{aligned}$$

The likelihood function for the ordinal logit model is,

$$\begin{aligned} L(\beta, \gamma; y) &= \prod_{i=1}^n \prod_{j=1}^J Pr(y_i = j | \beta, \gamma)^{I(y_i=j)} \\ &= \prod_{i=1}^n \prod_{j=1}^J \left[ \frac{e^{x_i' \beta} (e^{-\gamma_{j-1}} - e^{-\gamma_j})}{(1 + e^{-(\gamma_j - x_i' \beta)})(1 + e^{-(\gamma_{j-1} - x_i' \beta)})} \right]^{I(y_i=j)} \end{aligned}$$