## Advanced Econometrics PhD in Economics Exercise Sheet 4 - Limited Dependent Variable Models

1. Let  $t_i^*$  denote the duration of some event, such as unemployment, measured in continuous time. Consider the following model for  $t_i^*$ :

$$\begin{aligned} t_i^* &= \exp\left(\mathbf{X}_i'\beta_0 + u_i\right), \ u_i | X_i \sim N(0, \sigma^2), \\ t_i &= \min\left\{t_i^*, c\right\} \\ &= \begin{cases} t_i^* &, \text{ if } t_i^* \leq c \\ c &, \text{ if } t_i^* > c \end{cases}, \end{aligned}$$

where c > 0 is a known censoring constant.

- (a) Find  $P(t_i = c | X_i)$ , that is, the probability that the duration is censored. What happens as  $c \to \infty$ ?
- (b) What is the density of  $\log(t_i)$  (given  $\mathbf{X}_i$ ) when  $t_i < c$ ? Now write down the full density of  $\log(t_i)$  (given  $\mathbf{X}_i$ ).
- (c) Write down the log-likelihood function for observation i.
- (d) Partition  $\beta_0$  into the  $K_1$  and  $K_2$  sub-vectors  $\beta_{0,1}$  and  $\beta_{0,2}$ . Explain briefly how to test  $H_0$ :  $\beta_{0,2} = 0$ .
- (e) Obtain the log-likelihood function if the censoring time is potentially different for each person, so that  $t_i = \min\{t_i^*, c_i\}$ , where  $c_i$  is observed for all *i*. Assume that  $u_i$  is independent of  $(\mathbf{X}_i, c_i)$ .
- 2. Suppose that, for a random draw  $(Y_i, X_i)$  from the population,  $Y_i$  is a doubly censored variable:

$$Y_i = \min \{ \max \{a_1, Y_i^*\}, a_2 \}$$
  
= 
$$\begin{cases} a_1 & Y_i^* < a_1 \\ Y_i^* & a_1 \le Y_i^* \le a_2 \\ a_2 & Y_i^* > a_2 \end{cases},$$

where  $a_1 < a_2$  and  $Y_i^* | X_i \sim N(\mathbf{X}_i' \boldsymbol{\beta}_0, \sigma^2)$ .

- (a) Find  $P(Y_i = a_1 | \mathbf{X}_i)$  and  $P(Y_i = a_2 | \mathbf{X}_i)$  in terms of the standard normal cumulative distribution function,  $\mathbf{X}_i$ ,  $\boldsymbol{\beta}_0$ , and  $\sigma$ .
- (b) For  $y_i \in (a_1, a_2)$ , find  $P(Y_i \le y_i | \mathbf{X}_i)$  and use this to find the density function of  $Y_i$  given  $\mathbf{X}_i$ .
- (c) If  $z \sim N(0, 1)$  it can be shown that

$$E(z|c_1 < z < c_2) = \frac{\phi(c_1) - \phi(c_2)}{\Phi(c_1) - \Phi(c_2)}$$

for  $c_1 < c_2$ . Use this fact to find  $E(Y_i | a_1 < Y_i < a_2, \mathbf{X}_i)$  and  $E(Y_i | \mathbf{X}_i)$ 

- (d) Consider the following method for estimating  $\beta_0$ . Using only the uncensored observations, that is, observations for which  $a_1 < Y_i < a_2$ , run the OLS regression of  $Y_i$  on  $\mathbf{X}_i$ . Explain why this does not generally produce a consistent estimator of  $\beta_0$ .
- (e) Write down the log-likelihood function for observation i; it should consist of three parts.
- (f) For data censoring, how would the analysis change if  $a_1$  and  $a_2$  were replaced with  $a_{i1}$  and  $a_{i2}$ , respectively, where  $u_i$  is independent of  $(\mathbf{X}_i, a_{i1}, a_{i2})$ ?