

# TA session #6

## ECON 342

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February 25, 2009

**Problem 1.** Assume that you have a doubly censored variable, such that:

$$y_i^* = x_i\beta + \varepsilon_i$$
$$y_i = \begin{cases} a_1, & \text{if } y_i^* \leq a_1 \\ y_i^*, & \text{if } a_1 < y_i^* < a_2 \\ a_2, & \text{if } y_i^* \geq a_2 \end{cases}$$

- (a) Find  $\mathbb{P}(y_i = a_1|x_i)$ ,  $\mathbb{P}(y_i = a_2|x_i)$  and  $\mathbb{P}(y_i \leq z|x_i)$  for  $z \in (a_1, a_2)$ .
- (b) Find  $\mathbb{E}(y_i|x_i, a_1 < y_i < a_2)$  and  $\mathbb{E}(y_i|x_i)$ .
- (c) Consider the following method for estimating  $\beta$ . Using only the uncensored observations, run the OLS regression of  $y_i$  on  $x_i$ . Does this method produce a consistent estimator of  $\beta$ ?
- (d) What is the partial effect of  $x_{ij}$ ?
- (e) For data censoring, how would the analysis change if  $a_1$  and  $a_2$  were replaced with  $a_{11}$  and  $a_{12}$ ?

*Solution.* Using the definition of  $y_i$  we have,

$$\begin{aligned} \mathbb{P}(y_i = a_1|x_i) &= \mathbb{P}(y_i^* \leq a_1|x_i) \\ &= \mathbb{P}(\varepsilon_i \leq a_1 - x_i\beta|x_i) \\ &= \mathbb{P}\left(\frac{\varepsilon_i}{\sigma} \leq \frac{a_1 - x_i\beta}{\sigma} | x_i\right) \\ &= \Phi\left(\frac{a_1 - x_i\beta}{\sigma}\right) \end{aligned}$$

and,

$$\begin{aligned}
\mathbb{P}(y_i = a_2|x_i) &= \mathbb{P}(y_i^* \geq a_2|x_i) \\
&= \mathbb{P}(\varepsilon_i \geq a_2 - x_i\beta|x_i) \\
&= \mathbb{P}\left(\frac{\varepsilon_i}{\sigma} \geq \frac{a_2 - x_i\beta}{\sigma}|x_i\right) \\
&= 1 - \Phi\left(\frac{a_2 - x_i\beta}{\sigma}\right)
\end{aligned}$$

In the third case, note that when  $y_i = y_i^*$  for values of  $y_i^* \in (a_1, a_2)$ , then,

$$\mathbb{P}(y_i \leq z|x_i) = \mathbb{P}(y_i^* \leq z|x_i) = \Phi\left(\frac{z - x_i\beta}{\sigma}\right)$$

To obtain the conditional expectation  $\mathbb{E}(y_i|x_i, a_1 < y_i < a_2)$  from the equation above note,

$$f_y(z|x) = \mathbb{P}(y_i^* = z|x_i) = \frac{\partial \mathbb{P}(y_i^* \leq z|x_i)}{\partial z} = \frac{1}{\sigma} \phi\left(\frac{z - x_i\beta}{\sigma}\right)$$

then,

$$f_y(z|x, a_1 < z < a_2) = \mathbb{P}(y_i^* = z|x_i, a_1 < z < a_2) = \frac{f(z|x)}{F(a_2|x) - F(a_1|x)}$$

which using the above definitions equals,

$$f_y(z|x, a_1 < z < a_2) = \frac{\frac{1}{\sigma} \phi\left(\frac{z - x_i\beta}{\sigma}\right)}{\Phi\left(\frac{a_2 - x_i\beta}{\sigma}\right) - \Phi\left(\frac{a_1 - x_i\beta}{\sigma}\right)}$$

Then,

$$\mathbb{E}(y_i|x_i, a_1 < y_i < a_2) = \mathbb{E}(y_i^*|x_i, a_1 < y_i^* < a_2) = \int_{a_1}^{a_2} y_i^* \frac{\frac{1}{\sigma} \phi\left(\frac{y_i^* - x_i\beta}{\sigma}\right)}{\Phi\left(\frac{a_2 - x_i\beta}{\sigma}\right) - \Phi\left(\frac{a_1 - x_i\beta}{\sigma}\right)} dy_i^* \quad (1)$$

To solve the integral, re-write (1) as,

$$\mathbb{E}(y_i^*|x_i, a_1 < y_i^* < a_2) = \frac{1}{\Phi\left(\frac{a_2 - x_i\beta}{\sigma}\right) - \Phi\left(\frac{a_1 - x_i\beta}{\sigma}\right)} \int_{a_1}^{a_2} \frac{y_i^* - x_i\beta}{\sigma} \phi\left(\frac{y_i^* - x_i\beta}{\sigma}\right) + \frac{x_i\beta}{\sigma} \phi\left(\frac{y_i^* - x_i\beta}{\sigma}\right) dy_i^*$$

and let  $u = \frac{y_i^* - x_i\beta}{\sigma}$  thus  $\sigma du = dy_i^*$ , then:

$$\sigma \int_{\frac{a_1 - x_i\beta}{\sigma}}^{\frac{a_2 - x_i\beta}{\sigma}} u \phi(u) du = \sigma \phi\left(\frac{a_1 - x_i\beta}{\sigma}\right) - \sigma \phi\left(\frac{a_2 - x_i\beta}{\sigma}\right) \quad (2)$$

which follows by noting that  $d\phi(u)/du = -u\phi(u)$ . Similarly,

$$\frac{x_i\beta}{\sigma} \int_{a_1}^{a_2} \phi\left(\frac{y_i^* - x_i\beta}{\sigma}\right) dy_i^* = \frac{x_i\beta}{\sigma} \sigma \int_{\frac{a_1 - x_i\beta}{\sigma}}^{\frac{a_2 - x_i\beta}{\sigma}} \phi(u) du = x_i\beta \left[ \Phi\left(\frac{a_2 - x_i\beta}{\sigma}\right) - \Phi\left(\frac{a_1 - x_i\beta}{\sigma}\right) \right] \quad (3)$$

With (2) and (3) we obtain:

$$\begin{aligned}\mathbb{E}(y_i|x_i, a_1 < y_i < a_2) &= \frac{1}{\Phi\left(\frac{a_2 - x_i\beta}{\sigma}\right) - \Phi\left(\frac{a_1 - x_i\beta}{\sigma}\right)} \left\{ \sigma\phi\left(\frac{a_1 - x_i\beta}{\sigma}\right) - \sigma\phi\left(\frac{a_2 - x_i\beta}{\sigma}\right) \right. \\ &\quad \left. + x_i\beta \left[ \Phi\left(\frac{a_2 - x_i\beta}{\sigma}\right) - \Phi\left(\frac{a_1 - x_i\beta}{\sigma}\right) \right] \right\} \\ &= x_i\beta + \sigma \frac{\phi\left(\frac{a_1 - x_i\beta}{\sigma}\right) - \phi\left(\frac{a_2 - x_i\beta}{\sigma}\right)}{\Phi\left(\frac{a_2 - x_i\beta}{\sigma}\right) - \Phi\left(\frac{a_1 - x_i\beta}{\sigma}\right)}\end{aligned}$$

Then, we can calculate  $\mathbb{E}(y_i|x_i)$  as follows

$$\begin{aligned}\mathbb{E}(y_i|x_i) &= a_1\mathbb{P}(y_i = a_1|x_i) + a_2\mathbb{P}(y_i = a_2|x_i) + \mathbb{E}(y_i|x_i, a_1 < y_i < a_2)\mathbb{P}(a_1 < y_i < a_2|x_i) \\ &= a_1\Phi\left(\frac{a_1 - x_i\beta}{\sigma}\right) + a_2 \left[ 1 - \Phi\left(\frac{a_2 - x_i\beta}{\sigma}\right) \right] + \left[ \Phi\left(\frac{a_2 - x_i\beta}{\sigma}\right) - \Phi\left(\frac{a_1 - x_i\beta}{\sigma}\right) \right] x_i\beta \\ &\quad + \sigma \left[ \phi\left(\frac{a_1 - x_i\beta}{\sigma}\right) - \phi\left(\frac{a_2 - x_i\beta}{\sigma}\right) \right] \\ &\neq x_i\beta\end{aligned}$$

which also suggests that estimating  $\beta$  using OLS will not produce a consistent estimate. Similarly, we found that

$$\mathbb{E}(y_i|x_i, a_1 < y_i < a_2) \neq x_i\beta$$

which also suggests that running a regression using only uncensored observations will lead to an inconsistent estimate of  $\beta$ .

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