

# HW2 Econometrics 3

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## Problem 2

Problem 2. Censoring/Truncation. Greene (2007) analyzed the default behavior and monthly behavior of a large sample of credit card users (13,444).

### (2.1)

Estimate the following model

$$\log \text{spend} = \beta_1 + \beta_2 \ln \text{income} + \beta_3 \text{Age} + \beta_4 \text{Adepcnt} + \beta_5 \text{ownrent} + \varepsilon$$

Table 1: Regression output used to answer Problem 2

	<i>Dependent variable:</i>		
	LOGSPEND		NA
	<i>OLS</i>	<i>censored regression</i>	<i>Heckman selection</i>
	(1)	(2)	(3)
Ln_income	1.121*** (0.033)	1.117*** (0.033)	0.907*** (0.162)
AGE	-0.015*** (0.001)	-0.014*** (0.001)	-0.014*** (0.002)
ADEPCNT	-0.027** (0.011)	-0.027** (0.011)	0.016 (0.034)
OWNRENT	-0.203*** (0.030)	-0.201*** (0.030)	-0.281*** (0.065)
logSigma		0.296*** (0.007)	
Constant	-3.363*** (0.243)	-3.340*** (0.246)	-1.419 (1.458)
Observations	10,499	10,499	13,444
R <sup>2</sup>	0.105		0.105
Adjusted R <sup>2</sup>	0.104		0.104
Log Likelihood		-18,012.210	
Akaike Inf. Crit.		36,036.430	
Bayesian Inf. Crit.		36,079.980	
$\rho$			-0.608
Inverse Mills Ratio			-0.878 (0.646)
Residual Std. Error	1.330 (df = 10494)		
F Statistic	306.358*** (df = 4; 10494)		
<i>Note:</i>		*p<0.1; **p<0.05; ***p<0.01	

(2.1.a)

Using OLS. What is the effect of 10% increase in income on credit card expenditure?

- Since we are dealing with log-log we can simply multiply the parameter estimate on income by ten, which gives 11.2120776. So a 10% increase in income is estimated to increase credit card spending by 11.2120776%.

(2.1.b)

Using Censored regression. What is the effect of 10% increase in income on credit card expenditure?

We will need to employ a Censored (Tobit) Regression and calculate the Partial Effects.

The general formulation for the Tobit Model (Greene 7th. ed., pg 848):

$$\begin{aligned} y_i^* &= x_i' \beta + \varepsilon_i \\ y_i &= 0 \quad \begin{cases} \text{if } y_i^* \leq 0 \\ \text{if } y_i^* \geq 0 \end{cases} \\ y_i &= y_i^* \end{aligned}$$

The censored regression model is a generalisation of the standard Tobit model. The dependent variable can be either left-censored, right-censored, or both left-censored and right-censored, where the lower and/or upper limit of the dependent variable can be any number:

$$y_i^* = x_i' \beta + \varepsilon_i \tag{1}$$

$$y_i = \begin{cases} a & \text{if } y_i^* \leq a \\ y_i^* & \text{if } a < y_i^* < b \\ b & \text{if } y_i^* \geq b \end{cases} \tag{2}$$

Here  $a$  is the lower limit and  $b$  is the upper limit of the dependent variable. If  $a = -\infty$  or  $b = \infty$ , the dependent variable is not left-censored or right-censored, respectively.

Censored regression models (including the standard Tobit model) are usually estimated by the Maximum Likelihood (ML) method. Assuming that the disturbance term  $\varepsilon$  follows a normal distribution with mean 0 and variance  $\sigma^2$ , the log-likelihood function is

$$\begin{aligned} \log L = \sum_{i=1}^N & \left[ I_i^a \log \Phi \left( \frac{a - x_i' \beta}{\sigma} \right) + I_i^b \log \Phi \left( \frac{x_i' \beta - b}{\sigma} \right) \right. \\ & \left. + (1 - I_i^a - I_i^b) \left( \log \phi \left( \frac{y_i - x_i' \beta}{\sigma} \right) - \log \sigma \right) \right], \end{aligned} \tag{3}$$

where  $\phi(\cdot)$  and  $\Phi(\cdot)$  denote the probability density function and the cumulative distribution function, respectively, of the standard normal distribution, and  $I_i^a$  and  $I_i^b$  are indicator functions with

$$I_i^a = \begin{cases} 1 & \text{if } y_i = a \\ 0 & \text{if } y_i > a \end{cases} \tag{4}$$

$$I_i^b = \begin{cases} 1 & \text{if } y_i = b \\ 0 & \text{if } y_i < b \end{cases} \tag{5}$$

The log-likelihood function of the censored regression model~(3) can be maximised with respect to the parameter vector  $(\beta', \sigma)'$  using standard non-linear optimisation algorithms.

The proper Partial Effects formula:

$$\frac{\partial E[y|x]}{\partial x} = \beta \Pr ob[a < y^* < b]$$

Where I compute the partial effect at each observation and then compute the mean.

The parginal effect of Ln\_income on LOGSPEND is 1.1169911. Therefore, a 10% increase of income is estimated to increase credit card spending by 11.169911.

### **(2.1.c)**

Using Heckman Two-Step Estimator. What the is effect of 10% increase in income on credit card expenditure?

11.240879