

Introduction to Econometrics
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Lecture 63

Qualitative Response Models - Probit and Tobit Models Part - 3

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Therefore we need consider the entire sample for estimating the Tobit model. we have the following two sets of observations:

1. some individuals with $y_i > 0$
 $y_i = \beta x_i + u_i$
 $\Rightarrow y_i - \beta x_i = u_i$
 $\Rightarrow \left(\frac{y_i - \beta x_i}{\sigma} \right) = \left(\frac{u_i}{\sigma} \right)$
 $f\left(\frac{y_i - \beta x_i}{\sigma} \right) \Rightarrow$ standard normal distn.
2. some individuals with y_i not observed, i.e. $y_i = 0$
for this group, what we know is that $y_i^* \leq 0$
 $\Rightarrow \beta x_i + u_i \leq 0$
 $\Rightarrow \left(\frac{u_i}{\sigma} \right) \leq - \left(\frac{\beta x_i}{\sigma} \right)$

$$Pr \left[\frac{u_i}{\sigma} \leq - \left(\frac{\beta x_i}{\sigma} \right) \right] = F \left[- \frac{\beta x_i}{\sigma} \right]$$


So, that means we need to consider the entire sample for estimating the Tobit model. Now, how will you proceed? So, if you look we have two set of observations. First of all, some individuals with y_i greater than 0. And when y_i is greater than 0 positive expenditure, so that means we can easily observe the function which is y_i equals to beta x_i plus u_i and then that implies y_i minus beta x_i equals to u_i or y_i minus beta x_i by sigma equals to u_i divided by sigma, this happens.


And probability of observing those households with y_i greater than 0 can be represented by the standard normal distribution. So, that means, probability of y_i greater than this can be represented by a standard normal distribution, which is y_i minus beta x_i by this. And the second set of observation, some individuals with y_i is, y_i not observed. That means, y_i not observed meaning y_i equals to 0.

For these set of people, for this group, what we know is that y_i start less than equals to 0, that means or you can say that beta, this implies beta x_i plus u_i less than equals to 0, that means u_i less than equals to minus beta x_i or we can say that by, u_i by sigma less than equals to beta x_i by sigma. And what is the probability of this? Probability that means, in terms of probability,

probability u_i by σ less than equals to minus βx_i by σ is actually F of minus βx_i by σ that is what we know.

So, that means, now, what do we need to construct like the Logit and Probit model a likelihood of observing the entire sample that means, within that sample some individuals with 0 observation and how do you them in terms of probability that can be represented as a standard normal distribution because when y_i greater than 0, this is nothing but y_i equals to $\beta x_i + u_i$. So, after some manipulation and dividing both sides by σ , you can say that F of y_i minus βx_i by this, this is basically a standard normal distribution and this is basically the cumulative standard normal distribution.

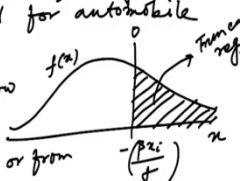
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$$\max_{(\beta, \sigma)} L = \prod_{\substack{y_i^* > 0 \\ y_i > 0}} f\left(\frac{y_i - \beta x_i}{\sigma}\right) \cdot \prod_{\substack{y_i^* \leq 0 \\ y_i = 0}} F\left(\frac{-\beta x_i}{\sigma}\right)$$

Tobit model is also known as

- ① Tobin's Probit model
 - since a Probit model is inherent in the Tobit as a first stage where individuals decide whether to buy a car or not
 - in the second stage they decide how much to spend for automobile
- ② Censored Regr.
 - censoring from below at zero
 - censoring is also possible from top or from both



So, if you combine these two, then you can actually write your log likelihood as some individuals for which y_i star greater than 0, what you can write is y_i minus βx_i by σ and then multiplied by then for some household for which y_i star less than equals to 0 that is nothing but minus βx_i by σ .

So, this is your likelihood function which you are trying to maximize and then with respect to β and σ . And if you maximize this likelihood, you will estimate your β hat and σ hat by maximum likelihood method. Now, detailed procedure of again deriving the log likelihood and other things I am not going in detail about that. What I am trying to make you understand is how intuitively you can understand the Tobit model, wherein my objective is to estimate the elasticity.

That means, I am not removing those individuals with 0 value for the automobile expenditure rather I am trying to construct a likelihood with both sets of people, for the first set of people y_i star greater than 0 or y_i greater than 0, you have positive automobile expenditure and remaining is y_i equals to 0 because y_i star less than or equals to 0. So, here y_i also greater than 0, here y_i equals to 0, this is one.

This is the second set of people, this is the first set of people and probability of observing the entire sample is then the likelihood. And how will you maximize this likelihood? If you maximize this likelihood with respect to beta and sigma, then again you will get the beta hat and sigma hat which maximizes the likelihood of observing a particular sample.

Now, this Tobit model is basically known as Tobin's Probit model, because in Tobit model there is a Probit model also which is inbuilt, why this is so? Because if you think closely to observe the positive expenditure for automobile, that particular household or individual, firstly have to decide that yes, I will buy a car and then the next question is how much will you spend.

If you do not decide to buy the car, I am not able to observe your automobile expenditure that means, you can think of the Tobit model in terms of two different steps. In the first step the individual is deciding whether to buy a car or not, which is exactly like the Probit model probability of buying a car. If I pass the first stage that means, if probability of y_i equals to 1 in the first stage, then in the next stage what would be my expenditure?

That is why since the Probit model is inbuilt in the Tobit model as a first stage, the Tobit model is known as Tobin's Probit model. Why this is so? Since, a Probit model is inbuilt in the Tobit as a first stage, where individuals decide whether to buy a car or not. And in the second stage they decide how much to spend for automobile. And secondly, this is also known as a censored model, censored regression. Why this is censored regression?

Because if you look at the diagram once again this is x , this is $f(x)$ and this is minus beta x_i by sigma. So, that means, I can observe a positive expenditure only after this. So, that means I am putting a censoring at 0. So, 0 or less than that I am not able to observe, if your expenditure is more than 0, then only I will be able to observe your automobile expenditure. That is why this is called a censored regression.

Now you can do censoring from the below or upper limit also. For example, let us say I am saying you are observing the efficiency of a firm, where efficiency is defined as your actual

output to potential output. And in that case, the maximum of the technical efficiency of a firm value would be just 1 where actual and potential is same.

In that case, any value of technical efficiency which is more than 1 is not observable, it is meaningless. So, that means technical efficiency should be censored from top. What would be the censoring point? So, this is a censoring at 0. So, that is why this is censoring from below at 0. Censoring is also possible from top or from both.

So, in this particular automobile expenditure case, we can say that we will observe the individual also automobile expenditure if it is greater than 0 and there is no upper limit. So, in that case, it is known as censoring from the below or lower censoring or left censoring.

Now, a closely related concept is called truncated regression. what we are discussing earlier that for, if you remove this set of people with 0 or less than 0 expenditure and try to construct the model using only those individuals who have positive expenditure then here you have to apply truncated regression. u_i follows a truncated normal distribution. So, that means there is a difference between censoring and truncation and what is the difference?

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- Difference between censoring and Truncation
1. In censoring, we have limited data on y_i but complete information on x_i . In Truncation, we have complete information on both y_i & x_i .
 2. While censoring is after sampling phenomena, Truncation is a pre-sampling decision.
- Truncated Regn example: Our objective is to estimate income elasticity of demand for rice for MGNREGA laborer.



The first difference is in censoring we have limited data on y_i which is a dependent variable, but complete information on x_i , because even if you do not buy a car, I can always observe your income. So, that is why irrespective of whether you buy a car or do not buy a car, I can always observe your income, but unless you buy a car, I cannot observe your automobile expenditure. That is why in censoring we have limited data on y_i but complete information on x_i . But in truncation, we have complete information on both y_i and x_i because I have already decided not to take those into my sample. This is the first difference.

The second one is while censoring is after sampling phenomena, truncation is a pre-sampling. So, that means, when our objective is to let us say find the automobile expenditure of the car owner, in the process of collecting the sample itself, I will not ask you the question unless and until I see you have a car. So, my interest is now finding the automobile expenditure of the car or not, but what was happening in case of censoring? I am going and asking each and every individual what is your expenditure for automobile and you will respond either with actual expenditure or 0 value.

After collecting the sample, I will now put a censoring on 0. I will say that no these are the samples, not households with 0 observation, 0 expenditure. So, I will put a 0 expenditure here and then I will estimate the Tobit model. That is why while censoring is actually after sampling phenomena, truncation is a pre-sampling decision.

So, these are the 2 basic difference between censoring and truncation we need to keep in mind and if at all you want to remove those observations and consider only the selected

sample of car or not to estimate this, that means, you need to apply a specific technique of estimation which is called truncated regression, which is quite different from the Tobit model estimation using the likelihood method.

So, this is basically, these are the two differences between Tobit regression which is also known as limited dependent variable or censored regression model and this is a truncated regression model. We are not discussing the truncated regression here, we will discuss only the Tobit model or limited dependent variable or censored regressive model, but you should always keep this thing in mind.

Because sometimes your objective might be such that you have to apply only the truncated regression. For example, if your objective is let us say is to estimate income elasticity of demand for rice for MGNREGA laborer. That means, I am not going to ask each and every individual how much do you spend for rice? Rather I will focus only those individual who are MGNREGA laborer. So, that means, while collecting the sample itself it is truncated. I will take only those who are MGNREGA people. Like the previous case, I will ask only those people who have, for example, here if your income is not coming from MGNREGA, then you are not, I am not considering you. That is why income elasticity of demand for rice only for MGNREGA people.

So, there is no question of considering the entire sample, which is basically a truncated regression here. So, with this, we are just closing our discussion today. And tomorrow, what we will do, we will discuss about the estimation part of the Tobit model and we will also see how to interpret the coefficients. Thank you.