

**Econ 240B Spring 2009**  
**Problem Set 4**

**This problem set is due in class on Monday May 4th, 2009**

1. Derive the Multinomial Logit Model. Define the latent utility of choosing alternative  $m$ ,  $m = 1, \dots, M$ , to be:

$$y_{im}^* = x_i' \beta_m + \epsilon_{im}$$

where conditional on  $x_{im}$ ,  $m = 1, \dots, M$ ,  $\epsilon_{im}$ ,  $m = 1, \dots, M$  are independently and identically distributed with type I extreme value distribution function:  $F(\epsilon) = \exp(-\exp(-\epsilon))$ . Individual  $i$  chooses alternative  $m$  if and only if it yields the highest latent utility, i.e.

$$y_{im} = \begin{cases} 1 & \text{if } y_{im}^* \geq y_{im'}^*, \forall m' \neq m, m' = 1, \dots, M \\ 0 & \text{otherwise} \end{cases}$$

Show that

$$P(y_{im} = 1 | x_i) = \frac{\exp(x_i' \beta_m)}{\sum_{m'=1}^M \exp(x_i' \beta_{m'})}$$

How do you interpret the coefficients  $\beta_m$ ?

You may take  $M = 3$  if you find it notationally cumbersome to work with general  $M$ .

2. Answer the following questions as true, false or uncertain and explain your answers.

(a) consider the limited dependent variable model

$$y(t) = \begin{cases} x(t) \beta + u(t) & \text{if } x(t) \beta + u(t) > 0 \\ 0 & \text{otherwise} \end{cases}$$

Regressing  $y(t)$  on  $x(t)$  using only those observations for which  $y(t) \neq 0$  yields estimates that overstates the true value of  $\beta$ .

(b) Consider the limited dependent variable model

$$y(t) = \begin{cases} x(t) \beta + u(t) & \text{if } x(t) \gamma + e(t) > 0 \\ 0 & \text{otherwise} \end{cases}$$

where  $(u(t), e(t)) \sim N(0, \Sigma)$ . Applying nonlinear least square to the regression equation of  $y(t)$  on  $x(t)$  for the sample of observation with  $y(t) \neq 0$ :

$$y(t) = x(t) \beta + \rho \frac{\phi(x(t) \gamma)}{\Phi(x(t) \gamma)} + \eta(t)$$

will yield consistent estimate of the parameter  $\beta$ .  $\phi$  and  $\Phi$  are the density and the c.d.f. of a standardized normal distribution, and  $\rho$  is a parameter.

- (c) The answer to question (2.b) does not change if merely  $E(u(t) | e(t)) = \rho e(t)$  where  $e(t) \sim N(0, 1)$ .

(d) Consider the probit model  $\delta(t) = 1$  when  $y(t) = x(t)\beta + e(t) > 0$  with  $e(t) \sim iid N(0, 1)$ , and  $\delta(t) = 0$  otherwise. Applying weighted nonlinear LS to the regression equation  $\delta(t) = 1 - \Phi(-x(t)'\beta) + u(t)$  (where  $\Phi$  is the standard normal cdf) yields a consistent estimator for  $\beta$  that is as efficient as the maximum likelihood estimator for  $\beta$ .

3. This is an empirical exercise about multinomial logit models.

Use the data set `bell_female.dat`. Consider a three-state classification of a woman's hours of work: she doesn't work at all (designate by setting the discrete variable  $b=1$ ); she works part of the year which implies that  $0 < \text{Weeks} < 20$  (designated by  $b=2$ ); and she works most of the year with  $\text{Weeks} \geq 20$  ( $b=3$ ). Estimate the probability  $Pr(b = j|x)$  using a multinomial logit model and test whether marriage influences the likelihood that a woman works part of the year instead of most of the year.