

**Example 6:** Nested logit model:

- (i) In the 1st step, choose YES or NO. Each probability is  $P_Y$  and  $P_N = 1 - P_Y$ .
- (ii) Stop if NO is chosen in the 1st step. Go to the next if YES is chosen in the 1st step.
- (iii) In the 2nd step, choose A or B if YES is chosen in the 1st step. Each probability is  $P_{A|Y}$  and  $P_{B|Y}$ .

For simplicity, usually we assume the logistic distribution.

So, we call the nested logit model.

The probability that the  $i$ th individual chooses NO is:

$$P_{N,i} = \frac{1}{1 + \exp(X_i\beta)}.$$

The probability that the  $i$ th individual chooses YES and A is:

$$P_{A|Y,i}P_{Y,i} = P_{A|Y,i}(1 - P_{N,i}) = \frac{\exp(Z_i\alpha)}{1 + \exp(Z_i\alpha)} \frac{\exp(X_i\beta)}{1 + \exp(X_i\beta)}.$$

The probability that the  $i$ th individual chooses YES and B is:

$$P_{B|Y,i}P_{Y,i} = (1 - P_{A|Y,i})(1 - P_{N,i}) = \frac{1}{1 + \exp(Z_i\alpha)} \frac{\exp(X_i\beta)}{1 + \exp(X_i\beta)}.$$

In the 1st step, decide if the  $i$ th individual buys a car or not.

In the 2nd step, choose A or B.

$X_i$  includes annual income, distance from the nearest station, and so on.

$Z_i$  are speed, fuel-efficiency, car company, color, and so on.

The likelihood function is:

$$\begin{aligned} L(\alpha, \beta) &= \prod_{i=1}^n P_{N,i}^{I_{1i}} \left( ((1 - P_{N,i})P_{A|Y,i})^{I_{2i}} ((1 - P_{N,i})(1 - P_{A|Y,i}))^{1-I_{2i}} \right)^{1-I_{1i}} \\ &= \prod_{i=1}^n P_{N,i}^{I_{1i}} (1 - P_{N,i})^{1-I_{1i}} \left( P_{A|Y,i}^{I_{2i}} (1 - P_{A|Y,i})^{1-I_{2i}} \right)^{1-I_{1i}}, \end{aligned}$$

where

$$I_{1i} = \begin{cases} 1, & \text{if the } i\text{th individual decides not to buy a car in the 1st step,} \\ 0, & \text{if the } i\text{th individual decides to buy a car in the 1st step,} \end{cases}$$

$$I_{2i} = \begin{cases} 1, & \text{if the } i\text{th individual chooses A in the 2nd step,} \\ 0, & \text{if the } i\text{th individual chooses B in the 2nd step,} \end{cases}$$

Remember that  $E(y_i) = F(X_i\beta^*)$ , where  $\beta^* = \frac{\beta}{\sigma}$ .

Therefore, size of  $\beta^*$  does not mean anything.

The marginal effect is given by:

$$\frac{\partial E(y_i)}{\partial X_i} = f(X_i\beta^*)\beta^*.$$

Thus, the marginal effect depends on the height of the density function  $f(X_i\beta^*)$ .

## 7.2 Limited Dependent Variable Model (制限従属変数モデル)

**Truncated Regression Model:** Consider the following model:

$$y_i = X_i\beta + u_i, \quad u_i \sim N(0, \sigma^2) \text{ when } y_i > a, \text{ where } a \text{ is a constant,}$$

for  $i = 1, 2, \dots, n$ .

Consider the case of  $y_i > a$  (i.e., in the case of  $y_i \leq a$ ,  $y_i$  is not observed).

$$E(u_i | X_i\beta + u_i > a) = \int_{a - X_i\beta}^{\infty} u_i \frac{f(u_i)}{1 - F(a - X_i\beta)} du_i.$$

Suppose that  $u_i \sim N(0, \sigma^2)$ , i.e.,  $\frac{u_i}{\sigma} \sim N(0, 1)$ .

Using the following standard normal density and distribution functions:

$$\begin{aligned}\phi(x) &= (2\pi)^{-1/2} \exp\left(-\frac{1}{2}x^2\right), \\ \Phi(x) &= \int_{-\infty}^x (2\pi)^{-1/2} \exp\left(-\frac{1}{2}z^2\right) dz = \int_{-\infty}^x \phi(z) dz,\end{aligned}$$

$f(x)$  and  $F(x)$  are given by:

$$f(x) = (2\pi\sigma^2)^{-1/2} \exp\left(-\frac{1}{2\sigma^2}x^2\right) = \frac{1}{\sigma}\phi\left(\frac{x}{\sigma}\right),$$
$$F(x) = \int_{-\infty}^x (2\pi\sigma^2)^{-1/2} \exp\left(-\frac{1}{2\sigma^2}z^2\right)dz = \Phi\left(\frac{x}{\sigma}\right).$$

**[Review — Mean of Truncated Normal Random Variable:]**

Let  $X$  be a normal random variable with mean  $\mu$  and variance  $\sigma^2$ .

Consider  $E(X|X > a)$ , where  $a$  is known.

The truncated distribution of  $X$  given  $X > a$  is:

$$f(x|x > a) = \frac{(2\pi\sigma^2)^{-1/2} \exp\left(-\frac{1}{2\sigma^2}(x - \mu)^2\right)}{\int_a^{\infty} (2\pi\sigma^2)^{-1/2} \exp\left(-\frac{1}{2\sigma^2}(x - \mu)^2\right)dx} = \frac{\frac{1}{\sigma}\phi\left(\frac{x}{\sigma}\right)}{1 - \Phi\left(\frac{a - \mu}{\sigma}\right)}.$$

$$\begin{aligned}
E(X|X > a) &= \int_a^{\infty} x f(x|x > a) dx = \frac{\int_a^{\infty} x (2\pi\sigma^2)^{-1/2} \exp\left(-\frac{1}{2\sigma^2}(x - \mu)^2\right) dx}{\int_a^{\infty} (2\pi\sigma^2)^{-1/2} \exp\left(-\frac{1}{2\sigma^2}(x - \mu)^2\right) dx} \\
&= \frac{\sigma\phi\left(\frac{a - \mu}{\sigma}\right) + \mu\left(1 - \Phi\left(\frac{a - \mu}{\sigma}\right)\right)}{1 - \Phi\left(\frac{a - \mu}{\sigma}\right)} = \frac{\sigma\phi\left(\frac{a - \mu}{\sigma}\right)}{1 - \Phi\left(\frac{a - \mu}{\sigma}\right)} + \mu,
\end{aligned}$$

which are shown below. The denominator is:

$$\begin{aligned}
\int_a^{\infty} (2\pi\sigma^2)^{-1/2} \exp\left(-\frac{1}{2\sigma^2}(x - \mu)^2\right) dx &= \int_{\frac{a - \mu}{\sigma}}^{\infty} (2\pi)^{-1/2} \exp\left(-\frac{1}{2}z^2\right) dz \\
&= 1 - \int_{-\infty}^{\frac{a - \mu}{\sigma}} (2\pi)^{-1/2} \exp\left(-\frac{1}{2}z^2\right) dz \\
&= 1 - \Phi\left(\frac{a - \mu}{\sigma}\right),
\end{aligned}$$

where  $x$  is transformed into  $z = \frac{x - \mu}{\sigma}$ .  $x > a \implies z = \frac{x - \mu}{\sigma} > \frac{a - \mu}{\sigma}$ .

The numerator is:

$$\begin{aligned}
 & \int_a^\infty x(2\pi\sigma^2)^{-1/2} \exp(-\frac{1}{2\sigma^2}(x-\mu)^2)dx \\
 &= \int_{\frac{a-\mu}{\sigma}}^\infty (\sigma z + \mu)(2\pi)^{-1/2} \exp(-\frac{1}{2}z^2)dz \\
 &= \sigma \int_{\frac{a-\mu}{\sigma}}^\infty z(2\pi)^{-1/2} \exp(-\frac{1}{2}z^2)dz + \mu \int_{\frac{a-\mu}{\sigma}}^\infty (2\pi)^{-1/2} \exp(-\frac{1}{2\sigma^2}z^2)dz \\
 &= \sigma \int_{\frac{1}{2}(\frac{a-\mu}{\sigma})^2}^\infty (2\pi)^{-1/2} \exp(-t)dt + \mu(1 - \Phi(\frac{a-\mu}{\sigma})) \\
 &= \sigma\phi(\frac{a-\mu}{\sigma}) + \mu(1 - \Phi(\frac{a-\mu}{\sigma})),
 \end{aligned}$$

where  $z$  is transformed into  $t = \frac{1}{2}z^2$ .  $z > \frac{a-\mu}{\sigma} \implies t = \frac{1}{2}z^2 > \frac{1}{2}(\frac{a-\mu}{\sigma})^2$ .

**[End of Review]**

Therefore, the conditional expectation of  $u_i$  given  $X_i\beta + u_i > a$  is:

$$\begin{aligned} E(u_i|X_i\beta + u_i > a) &= \int_{a-X_i\beta}^{\infty} u_i \frac{f(u_i)}{1 - F(a - X_i\beta)} du_i = \int_{a-X_i\beta}^{\infty} \frac{u_i}{\sigma} \frac{\phi(\frac{u_i}{\sigma})}{1 - \Phi(\frac{a - X_i\beta}{\sigma})} du_i \\ &= \frac{\sigma \phi(\frac{a - X_i\beta}{\sigma})}{1 - \Phi(\frac{a - X_i\beta}{\sigma})}. \end{aligned}$$

Accordingly, the conditional expectation of  $y_i$  given  $y_i > a$  is given by:

$$\begin{aligned} E(y_i|y_i > a) &= E(y_i|X_i\beta + u_i > a) = E(X_i\beta + u_i|X_i\beta + u_i > a) \\ &= X_i\beta + E(u_i|X_i\beta + u_i > a) = X_i\beta + \frac{\sigma \phi(\frac{a - X_i\beta}{\sigma})}{1 - \Phi(\frac{a - X_i\beta}{\sigma})}, \end{aligned}$$

for  $i = 1, 2, \dots, n$ .

## Estimation:

MLE:

$$L(\beta, \sigma^2) = \prod_{i=1}^n \frac{f(y_i - X_i\beta)}{1 - F(a - X_i\beta)} = \prod_{i=1}^n \frac{1}{\sigma} \frac{\phi(\frac{y_i - X_i\beta}{\sigma})}{1 - \Phi(\frac{a - X_i\beta}{\sigma})}$$

is maximized with respect to  $\beta$  and  $\sigma^2$ .

## Some Examples:

### 1. Buying a Car:

$y_i = x_i\beta + u_i$ , where  $y_i$  denotes expenditure for a car, and  $x_i$  includes income, price of the car, etc.

Data on people who bought a car are observed.

People who did not buy a car are ignored.

## 2. Working-hours of Wife:

$y_i$  represents working-hours of wife, and  $x_i$  includes the number of children, age, education, income of husband, etc.

## 3. Stochastic Frontier Model:

$y_i = f(K_i, L_i) + u_i$ , where  $y_i$  denotes production,  $K_i$  is stock, and  $L_i$  is amount of labor.

We always have  $y_i \leq f(K_i, L_i)$ , i.e.,  $u_i \leq 0$ .

$f(K_i, L_i)$  is a maximum value when we input  $K_i$  and  $L_i$ .

## Censored Regression Model or Tobit Model:

$$y_i = \begin{cases} X_i\beta + u_i, & \text{if } y_i > a, \\ a, & \text{otherwise.} \end{cases}$$

The probability which  $y_i$  takes  $a$  is given by:

$$P(y_i = a) = P(y_i \leq a) = F(a) \equiv \int_{-\infty}^a f(x)dx,$$

where  $f(\cdot)$  and  $F(\cdot)$  denote the density function and cumulative distribution function of  $y_i$ , respectively.

Therefore, the likelihood function is:

$$L(\beta, \sigma^2) = \prod_{i=1}^n F(a)^{I(y_i=a)} \times f(y_i)^{1-I(y_i=a)},$$

where  $I(y_i = a)$  denotes the indicator function which takes one when  $y_i = a$  or zero otherwise.

When  $u_i \sim N(0, \sigma^2)$ , the likelihood function is:

$$L(\beta, \sigma^2) = \prod_{i=1}^n \left( \int_{-\infty}^a (2\pi\sigma^2)^{-1/2} \exp\left(-\frac{1}{2\sigma^2}(y_i - X_i\beta)^2\right) dy_i \right)^{I(y_i=a)} \\ \times \left( (2\pi\sigma^2)^{-1/2} \exp\left(-\frac{1}{2\sigma^2}(y_i - X_i\beta)^2\right) \right)^{1-I(y_i=a)},$$

which is maximized with respect to  $\beta$  and  $\sigma^2$ .

## Example of Truncated Regression Model:

### Demand Function of Watermelon

二人以上の世帯のうち勤労者世帯（2000 年～）

y 実収入【円】  
 a りんご【円】  
 g ぶどう【円】  
 w すいか【円】  
 ag りんご【1g】  
 gg ぶどう【1g】  
 wg すいか【1g】  
 CPI CPI

| year    | y      | a   | g    | w   | ag   | gg   | wg   | CPI   |
|---------|--------|-----|------|-----|------|------|------|-------|
| 2000.01 | 458911 | 371 | 6    | 3   | 1093 | 9    | 3    | 102.8 |
| 2000.02 | 486601 | 416 | 4    | 4   | 1285 | 5    | 4    | 102.5 |
| 2000.03 | 494395 | 388 | 8    | 8   | 1145 | 12   | 10   | 102.7 |
| 2000.04 | 505409 | 350 | 19   | 46  | 899  | 28   | 85   | 102.9 |
| 2000.05 | 460116 | 258 | 46   | 243 | 598  | 45   | 638  | 103.0 |
| 2000.06 | 772611 | 191 | 153  | 352 | 446  | 169  | 1163 | 102.8 |
| 2000.07 | 640258 | 139 | 317  | 571 | 306  | 293  | 2152 | 102.5 |
| 2000.08 | 506757 | 144 | 1032 | 397 | 282  | 1073 | 1558 | 102.8 |
| 2000.09 | 446405 | 354 | 826  | 30  | 884  | 1002 | 109  | 102.7 |
| 2000.10 | 488921 | 501 | 292  | 3   | 1460 | 360  | 8    | 102.7 |
| 2000.11 | 457054 | 739 | 37   | 1   | 2024 | 43   | 2    | 102.4 |

|         |         |     |     |     |      |     |      |       |
|---------|---------|-----|-----|-----|------|-----|------|-------|
| 2000.12 | 1035616 | 938 | 16  | 5   | 2230 | 27  | 11   | 102.5 |
| 2001.01 | 453748  | 329 | 11  | 1   | 905  | 16  | 2    | 102.5 |
| 2001.02 | 475556  | 350 | 5   | 1   | 920  | 6   | 3    | 102.1 |
| 2001.03 | 481198  | 321 | 7   | 3   | 835  | 11  | 5    | 101.9 |
| 2001.04 | 498080  | 287 | 17  | 52  | 713  | 26  | 92   | 102.1 |
| 2001.05 | 447510  | 255 | 43  | 236 | 582  | 43  | 602  | 102.2 |
| 2001.06 | 766471  | 169 | 138 | 355 | 352  | 120 | 1167 | 101.9 |
| 2001.07 | 614715  | 108 | 301 | 616 | 203  | 278 | 2403 | 101.6 |
| 2001.08 | 496482  | 129 | 827 | 400 | 265  | 916 | 1577 | 102.0 |
| 2001.09 | 447397  | 449 | 661 | 26  | 1087 | 823 | 90   | 101.8 |
| 2001.10 | 489834  | 598 | 241 | 1   | 1581 | 308 | 2    | 101.8 |
| 2001.11 | 461094  | 673 | 27  | 1   | 2026 | 34  | 2    | 101.3 |
| 2001.12 | 1000728 | 961 | 16  | 3   | 2622 | 16  | 6    | 101.2 |
| 2002.01 | 462389  | 331 | 4   | 2   | 997  | 4   | 3    | 101.0 |
| 2002.02 | 477622  | 343 | 2   | 1   | 1327 | 2   | 1    | 100.5 |
| 2002.03 | 496351  | 326 | 8   | 8   | 1114 | 10  | 22   | 100.7 |
| 2002.04 | 485770  | 273 | 14  | 50  | 826  | 21  | 90   | 101.0 |
| 2002.05 | 444612  | 243 | 57  | 208 | 726  | 55  | 517  | 101.3 |
| 2002.06 | 745480  | 194 | 170 | 353 | 524  | 157 | 1225 | 101.2 |
| 2002.07 | 583862  | 126 | 324 | 499 | 313  | 341 | 2075 | 100.8 |
| 2002.08 | 488257  | 151 | 722 | 335 | 312  | 813 | 1406 | 101.1 |
| 2002.09 | 440319  | 376 | 730 | 24  | 939  | 853 | 88   | 101.1 |
| 2002.10 | 475494  | 506 | 366 | 1   | 1504 | 462 | 3    | 100.9 |
| 2002.11 | 439186  | 733 | 36  | 3   | 2056 | 52  | 3    | 100.9 |
| 2002.12 | 939747  | 847 | 24  | 2   | 2599 | 38  | 2    | 100.9 |
| 2003.01 | 435989  | 303 | 7   | 1   | 900  | 12  | 0    | 100.6 |
| 2003.02 | 455309  | 305 | 3   | 2   | 1148 | 5   | 1    | 100.3 |
| 2003.03 | 456873  | 326 | 11  | 2   | 1094 | 22  | 8    | 100.6 |
| 2003.04 | 475037  | 273 | 18  | 36  | 815  | 28  | 63   | 100.9 |

|         |        |     |     |     |      |     |      |       |
|---------|--------|-----|-----|-----|------|-----|------|-------|
| 2003.05 | 429669 | 221 | 40  | 171 | 583  | 58  | 422  | 101.1 |
| 2003.06 | 730617 | 157 | 177 | 294 | 368  | 150 | 967  | 100.8 |
| 2003.07 | 574574 | 153 | 244 | 379 | 326  | 242 | 1412 | 100.6 |
| 2003.08 | 474973 | 128 | 683 | 293 | 264  | 873 | 1110 | 100.8 |
| 2003.09 | 429301 | 333 | 636 | 33  | 938  | 738 | 88   | 100.9 |
| 2003.10 | 467408 | 506 | 258 | 5   | 1193 | 346 | 5    | 100.9 |
| 2003.11 | 435079 | 618 | 39  | 1   | 2105 | 46  | 0    | 100.4 |
| 2003.12 | 932887 | 757 | 12  | 3   | 1856 | 13  | 2    | 100.5 |
| 2004.01 | 445133 | 327 | 5   | 1   | 995  | 6   | 0    | 100.3 |
| 2004.02 | 474143 | 348 | 3   | 2   | 1044 | 4   | 3    | 100.3 |
| 2004.03 | 456288 | 287 | 7   | 5   | 829  | 9   | 6    | 100.5 |
| 2004.04 | 488217 | 221 | 13  | 52  | 640  | 26  | 114  | 100.5 |
| 2004.05 | 446758 | 192 | 52  | 168 | 487  | 61  | 542  | 100.6 |
| 2004.06 | 723370 | 141 | 133 | 289 | 362  | 123 | 725  | 100.8 |
| 2004.07 | 599045 | 94  | 313 | 462 | 223  | 307 | 1689 | 100.5 |
| 2004.08 | 476264 | 115 | 675 | 276 | 260  | 761 | 892  | 100.6 |
| 2004.09 | 440187 | 328 | 583 | 25  | 859  | 814 | 82   | 100.9 |
| 2004.10 | 467895 | 482 | 156 | 1   | 1192 | 204 | 4    | 101.4 |
| 2004.11 | 442885 | 563 | 48  | 2   | 1613 | 58  | 7    | 101.2 |
| 2004.12 | 920100 | 673 | 15  | 3   | 1686 | 24  | 0    | 100.7 |
| 2005.01 | 448635 | 310 | 6   | 4   | 785  | 9   | 3    | 100.5 |
| 2005.02 | 469673 | 340 | 4   | 6   | 911  | 4   | 17   | 100.2 |
| 2005.03 | 451360 | 360 | 11  | 7   | 933  | 13  | 9    | 100.5 |
| 2005.04 | 495036 | 294 | 18  | 23  | 787  | 30  | 37   | 100.6 |
| 2005.05 | 440388 | 226 | 47  | 149 | 485  | 50  | 416  | 100.7 |
| 2005.06 | 720667 | 152 | 126 | 337 | 335  | 111 | 1088 | 100.3 |
| 2005.07 | 576129 | 105 | 217 | 402 | 216  | 226 | 1546 | 100.2 |
| 2005.08 | 463034 | 104 | 582 | 328 | 234  | 652 | 1225 | 100.3 |
| 2005.09 | 427753 | 277 | 644 | 30  | 771  | 838 | 93   | 100.6 |

|         |        |     |     |     |      |     |      |       |
|---------|--------|-----|-----|-----|------|-----|------|-------|
| 2005.10 | 463838 | 404 | 363 | 1   | 1147 | 544 | 4    | 100.6 |
| 2005.11 | 433036 | 540 | 45  | 1   | 1594 | 67  | 1    | 100.2 |
| 2005.12 | 905473 | 631 | 13  | 2   | 1519 | 20  | 2    | 100.3 |
| 2006.01 | 437787 | 294 | 7   | 1   | 994  | 10  | 1    | 100.4 |
| 2006.02 | 461368 | 310 | 4   | 0   | 950  | 6   | 0    | 100.1 |
| 2006.03 | 429948 | 302 | 7   | 7   | 920  | 12  | 0    | 100.3 |
| 2006.04 | 472583 | 256 | 17  | 25  | 728  | 26  | 40   | 100.5 |
| 2006.05 | 426680 | 202 | 32  | 141 | 515  | 44  | 332  | 100.8 |
| 2006.06 | 684632 | 148 | 114 | 240 | 338  | 97  | 720  | 100.8 |
| 2006.07 | 613269 | 105 | 209 | 361 | 228  | 205 | 1413 | 100.5 |
| 2006.08 | 475866 | 82  | 595 | 324 | 163  | 634 | 1034 | 101.2 |
| 2006.09 | 429017 | 263 | 628 | 32  | 647  | 716 | 108  | 101.2 |
| 2006.10 | 467163 | 455 | 263 | 4   | 1144 | 359 | 4    | 101.0 |
| 2006.11 | 442147 | 605 | 23  | 0   | 1556 | 22  | 1    | 100.5 |
| 2006.12 | 968162 | 719 | 18  | 1   | 1949 | 13  | 0    | 100.6 |
| 2007.01 | 441039 | 309 | 5   | 1   | 858  | 4   | 0    | 100.4 |
| 2007.02 | 471681 | 319 | 3   | 8   | 950  | 5   | 0    | 99.9  |
| 2007.03 | 445076 | 346 | 6   | 2   | 1012 | 9   | 0    | 100.2 |
| 2007.04 | 472446 | 304 | 15  | 35  | 770  | 23  | 75   | 100.5 |
| 2007.05 | 431013 | 233 | 35  | 159 | 539  | 37  | 355  | 100.8 |
| 2007.06 | 735579 | 177 | 122 | 320 | 369  | 110 | 926  | 100.6 |
| 2007.07 | 592452 | 110 | 201 | 360 | 258  | 212 | 1322 | 100.5 |
| 2007.08 | 467786 | 103 | 581 | 341 | 211  | 639 | 1126 | 101.0 |
| 2007.09 | 431793 | 291 | 717 | 28  | 735  | 745 | 77   | 101.0 |
| 2007.10 | 469981 | 443 | 261 | 1   | 1185 | 331 | 3    | 101.3 |
| 2007.11 | 435640 | 574 | 45  | 0   | 1423 | 29  | 0    | 101.1 |
| 2007.12 | 950654 | 748 | 17  | 1   | 1873 | 27  | 0    | 101.3 |
| 2008.01 | 438998 | 302 | 4   | 2   | 835  | 5   | 0    | 101.1 |
| 2008.02 | 476282 | 309 | 4   | 0   | 884  | 5   | 0    | 100.9 |

|         |        |     |     |     |      |     |      |       |
|---------|--------|-----|-----|-----|------|-----|------|-------|
| 2008.03 | 453482 | 291 | 5   | 4   | 905  | 6   | 0    | 101.4 |
| 2008.04 | 469774 | 232 | 12  | 28  | 676  | 18  | 43   | 101.3 |
| 2008.05 | 435076 | 192 | 30  | 148 | 471  | 39  | 293  | 102.1 |
| 2008.06 | 737166 | 150 | 102 | 222 | 358  | 95  | 661  | 102.6 |
| 2008.07 | 587732 | 103 | 236 | 400 | 227  | 245 | 1212 | 102.8 |
| 2008.08 | 488216 | 88  | 615 | 307 | 197  | 670 | 1012 | 103.1 |
| 2008.09 | 433502 | 278 | 625 | 28  | 827  | 693 | 125  | 103.1 |
| 2008.10 | 481746 | 445 | 241 | 2   | 1336 | 337 | 7    | 103.0 |
| 2008.11 | 439394 | 526 | 36  | 0   | 1601 | 39  | 0    | 102.1 |
| 2008.12 | 969449 | 661 | 10  | 1   | 1949 | 13  | 2    | 101.7 |
| 2009.01 | 443337 | 268 | 5   | 0   | 865  | 17  | 0    | 101.1 |
| 2009.02 | 464665 | 277 | 3   | 1   | 1084 | 3   | 0    | 100.8 |
| 2009.03 | 443429 | 265 | 5   | 0   | 861  | 6   | 2    | 101.1 |
| 2009.04 | 473779 | 210 | 15  | 32  | 648  | 15  | 56   | 101.2 |
| 2009.05 | 436123 | 167 | 33  | 141 | 478  | 31  | 301  | 101.0 |
| 2009.06 | 700239 | 129 | 110 | 243 | 351  | 97  | 735  | 100.8 |
| 2009.07 | 573821 | 84  | 209 | 329 | 219  | 232 | 1253 | 100.5 |
| 2009.08 | 466393 | 80  | 493 | 303 | 193  | 494 | 1054 | 100.8 |
| 2009.09 | 422120 | 259 | 522 | 27  | 774  | 686 | 80   | 100.8 |
| 2009.10 | 459704 | 366 | 204 | 3   | 1129 | 248 | 5    | 100.4 |
| 2009.11 | 428219 | 558 | 41  | 1   | 1732 | 48  | 3    | 100.2 |
| 2009.12 | 906884 | 525 | 16  | 2   | 1561 | 17  | 2    | 100.0 |
| 2010.01 | 434344 | 256 | 7   | 0   | 804  | 5   | 0    | 100.1 |
| 2010.02 | 464866 | 265 | 2   | 0   | 917  | 3   | 0    | 100.0 |
| 2010.03 | 439410 | 264 | 5   | 4   | 829  | 8   | 10   | 100.3 |
| 2010.04 | 474616 | 208 | 12  | 11  | 578  | 21  | 19   | 100.4 |
| 2010.05 | 421413 | 167 | 31  | 102 | 391  | 31  | 219  | 100.3 |
| 2010.06 | 733886 | 129 | 96  | 205 | 285  | 93  | 513  | 100.1 |
| 2010.07 | 562094 | 78  | 183 | 339 | 168  | 161 | 1054 | 99.5  |

|         |        |     |     |     |      |     |      |       |
|---------|--------|-----|-----|-----|------|-----|------|-------|
| 2010.08 | 470717 | 67  | 543 | 327 | 141  | 566 | 935  | 99.7  |
| 2010.09 | 425771 | 245 | 608 | 22  | 567  | 624 | 36   | 99.9  |
| 2010.10 | 494398 | 371 | 237 | 2   | 955  | 271 | 5    | 100.2 |
| 2010.11 | 431281 | 541 | 44  | 1   | 1538 | 47  | 3    | 99.9  |
| 2010.12 | 895511 | 533 | 17  | 1   | 1511 | 23  | 0    | 99.6  |
| 2011.01 | 419728 | 239 | 6   | 0   | 666  | 6   | 0    | 99.5  |
| 2011.02 | 470071 | 257 | 6   | 0   | 732  | 6   | 0    | 99.5  |
| 2011.03 | 419862 | 250 | 8   | 0   | 758  | 13  | 0    | 99.8  |
| 2011.04 | 454433 | 210 | 16  | 19  | 634  | 27  | 25   | 99.9  |
| 2011.05 | 413506 | 177 | 37  | 115 | 508  | 54  | 281  | 99.9  |
| 2011.06 | 687212 | 158 | 84  | 206 | 416  | 70  | 606  | 99.7  |
| 2011.07 | 572662 | 97  | 162 | 351 | 257  | 138 | 849  | 99.7  |
| 2011.08 | 463760 | 94  | 487 | 285 | 204  | 508 | 909  | 99.9  |
| 2011.09 | 422720 | 230 | 453 | 35  | 621  | 517 | 136  | 99.9  |
| 2011.10 | 479749 | 350 | 215 | 3   | 932  | 220 | 0    | 100.0 |
| 2011.11 | 424272 | 410 | 41  | 1   | 1105 | 43  | 2    | 99.4  |
| 2011.12 | 893811 | 546 | 51  | 0   | 1487 | 67  | 0    | 99.4  |
| 2012.01 | 430477 | 252 | 7   | 0   | 574  | 12  | 0    | 99.6  |
| 2012.02 | 483625 | 268 | 7   | 0   | 647  | 8   | 0    | 99.8  |
| 2012.03 | 441015 | 257 | 16  | 2   | 505  | 21  | 1    | 100.3 |
| 2012.04 | 469381 | 199 | 25  | 19  | 355  | 42  | 30   | 100.4 |
| 2012.05 | 417723 | 158 | 38  | 99  | 312  | 51  | 145  | 100.1 |
| 2012.06 | 712592 | 129 | 90  | 181 | 208  | 87  | 567  | 99.6  |
| 2012.07 | 557032 | 97  | 166 | 326 | 179  | 129 | 1279 | 99.3  |
| 2012.08 | 470470 | 74  | 519 | 307 | 166  | 503 | 1087 | 99.4  |
| 2012.09 | 422046 | 211 | 491 | 44  | 513  | 567 | 118  | 99.6  |
| 2012.10 | 482101 | 355 | 295 | 2   | 945  | 342 | 3    | 99.6  |
| 2012.11 | 432681 | 482 | 50  | 1   | 1572 | 72  | 1    | 99.2  |
| 2012.12 | 902928 | 508 | 21  | 1   | 1404 | 29  | 0    | 99.3  |

|         |        |     |     |     |      |     |      |       |
|---------|--------|-----|-----|-----|------|-----|------|-------|
| 2013.01 | 433858 | 264 | 8   | 0   | 753  | 13  | 0    | 99.3  |
| 2013.02 | 476256 | 264 | 9   | 1   | 743  | 13  | 0    | 99.2  |
| 2013.03 | 444379 | 276 | 16  | 1   | 781  | 22  | 1    | 99.4  |
| 2013.04 | 479854 | 229 | 30  | 17  | 643  | 42  | 28   | 99.7  |
| 2013.05 | 422724 | 168 | 41  | 113 | 454  | 58  | 250  | 99.8  |
| 2013.06 | 728678 | 136 | 82  | 205 | 307  | 99  | 634  | 99.8  |
| 2013.07 | 569174 | 99  | 169 | 370 | 218  | 154 | 1204 | 100.0 |
| 2013.08 | 471411 | 75  | 480 | 284 | 182  | 506 | 862  | 100.3 |
| 2013.09 | 431931 | 217 | 566 | 27  | 544  | 621 | 85   | 100.6 |
| 2013.10 | 482684 | 374 | 231 | 2   | 1045 | 294 | 9    | 100.7 |
| 2013.11 | 436293 | 417 | 47  | 1   | 1080 | 56  | 0    | 100.8 |
| 2013.12 | 905822 | 574 | 25  | 0   | 1377 | 31  | 1    | 100.9 |
| 2014.01 | 438646 | 270 | 8   | 2   | 674  | 11  | 6    | 100.7 |
| 2014.02 | 479268 | 278 | 7   | 0   | 734  | 15  | 0    | 100.7 |
| 2014.03 | 438145 | 256 | 15  | 1   | 655  | 21  | 4    | 101.0 |
| 2014.04 | 463964 | 216 | 35  | 20  | 536  | 43  | 27   | 103.1 |
| 2014.05 | 421117 | 177 | 46  | 114 | 355  | 52  | 273  | 103.5 |
| 2014.06 | 710375 | 135 | 86  | 190 | 267  | 68  | 453  | 103.4 |
| 2014.07 | 555276 | 89  | 180 | 315 | 163  | 155 | 1067 | 103.4 |
| 2014.08 | 463810 | 82  | 511 | 224 | 147  | 431 | 704  | 103.7 |
| 2014.09 | 421809 | 236 | 528 | 34  | 574  | 551 | 147  | 103.9 |
| 2014.10 | 488273 | 379 | 250 | 2   | 1042 | 247 | 2    | 103.6 |
| 2014.11 | 431543 | 504 | 57  | 0   | 1397 | 57  | 1    | 103.2 |
| 2014.12 | 924911 | 692 | 22  | 0   | 1555 | 28  | 1    | 103.3 |
| 2015.01 | 440226 | 301 | 8   | 0   | 847  | 11  | 0    | 103.1 |
| 2015.02 | 488519 | 307 | 9   | 1   | 820  | 7   | 0    | 102.9 |
| 2015.03 | 449243 | 327 | 21  | 2   | 842  | 22  | 6    | 103.3 |
| 2015.04 | 476880 | 262 | 49  | 23  | 604  | 64  | 38   | 103.7 |
| 2015.05 | 430325 | 186 | 54  | 99  | 364  | 63  | 180  | 104.0 |

|         |        |     |     |     |      |     |      |       |
|---------|--------|-----|-----|-----|------|-----|------|-------|
| 2015.06 | 733589 | 140 | 96  | 203 | 235  | 80  | 529  | 103.8 |
| 2015.07 | 587156 | 101 | 189 | 297 | 158  | 146 | 1124 | 103.7 |
| 2015.08 | 475369 | 103 | 548 | 279 | 212  | 520 | 889  | 103.9 |
| 2015.09 | 415467 | 272 | 599 | 41  | 655  | 606 | 159  | 103.9 |
| 2015.10 | 485330 | 397 | 246 | 4   | 1107 | 252 | 19   | 103.9 |

```

-----
. gen t=_n                                <--- Make data t=1,2,...,190.

. tsset t
    time variable:  t, 1 to 190
        delta:      1 unit

. gen ry=log(y/(cpi/100))                  <--- log of real income

. gen rap=log((a/ag)/(cpi/100))           <--- log of real price of apple (yen/1g)

. gen rgp=log((g/gg)/(cpi/100))           <--- log of real price of grape (yen/1g)

. gen rwp=log((w/wg)/(cpi/100))           <--- log of real price of watermelon (yen/1g)
(40 missing values generated)

. gen lwg=log(wg)                         <--- log of demand of watermelon (1g)
(35 missing values generated)

. reg lwg ry rwp rap rgp if lwg>log(10)   <--- OLS using the data for wg>10

```

| Source |  | SS | df | MS | Number of obs = | 102 |
|--------|--|----|----|----|-----------------|-----|
|--------|--|----|----|----|-----------------|-----|

|          |            |     |            |
|----------|------------|-----|------------|
| Model    | 138.187919 | 4   | 34.5469797 |
| Residual | 83.2907458 | 97  | .858667482 |
| Total    | 221.478665 | 101 | 2.19285807 |

F( 4, 97) = 40.23  
 Prob > F = 0.0000  
 R-squared = 0.6239  
 Adj R-squared = 0.6084  
 Root MSE = .92664

| lwg   | Coef.     | Std. Err. | t     | P> t  | [95% Conf. Interval] |           |
|-------|-----------|-----------|-------|-------|----------------------|-----------|
| ry    | .8102279  | .5131024  | 1.58  | 0.118 | -.2081383            | 1.828594  |
| rwp   | -1.575157 | .3569839  | -4.41 | 0.000 | -2.283671            | -.8666422 |
| rap   | 2.854632  | .6983476  | 4.09  | 0.000 | 1.468606             | 4.240659  |
| rgp   | 2.158679  | .6110691  | 3.53  | 0.001 | .9458762             | 3.371482  |
| _cons | -3.826122 | 6.894227  | -0.55 | 0.580 | -17.50925            | 9.857011  |

. truncreg lwg ry rwp rap rgp if lwg>log(10) <--- This is equivalent to OLS  
 (note: 0 obs. truncated)

Fitting full model:

Iteration 0: log likelihood = -134.46068  
 Iteration 1: log likelihood = -134.39761  
 Iteration 2: log likelihood = -134.39733  
 Iteration 3: log likelihood = -134.39733

Truncated regression

Limit: lower = -inf  
 upper = +inf

Number of obs = 102  
 Wald chi2(4) = 169.23

Log likelihood = -134.39733

Prob > chi2 = 0.0000

| lwg    | Coef.     | Std. Err. | z     | P> z  | [95% Conf. Interval] |           |
|--------|-----------|-----------|-------|-------|----------------------|-----------|
| ry     | .8102279  | .5003683  | 1.62  | 0.105 | -.170476             | 1.790932  |
| rwp    | -1.575157 | .3481244  | -4.52 | 0.000 | -2.257468            | -.8928453 |
| rap    | 2.854632  | .6810162  | 4.19  | 0.000 | 1.519865             | 4.189399  |
| rgp    | 2.158679  | .5959037  | 3.62  | 0.000 | .9907293             | 3.326629  |
| _cons  | -3.826122 | 6.723128  | -0.57 | 0.569 | -17.00321            | 9.350967  |
| /sigma | .9036459  | .0632679  | 14.28 | 0.000 | .7796432             | 1.027649  |

. truncreg lwg ry rwp rap rgp if lwg>log(10), ll(log(10)) <--- truncated reg  
(note: 0 obs. truncated)

Fitting full model:

Iteration 0: log likelihood = -132.93358

Iteration 1: log likelihood = -132.70871

Iteration 2: log likelihood = -132.70789

Iteration 3: log likelihood = -132.70789

Truncated regression

Limit: lower = 2.3025851

upper = +inf

Log likelihood = -132.70789

Number of obs = 102

Wald chi2(4) = 145.68

Prob > chi2 = 0.0000

| lwg    | Coef.     | Std. Err. | z     | P> z  | [95% Conf. Interval] |          |
|--------|-----------|-----------|-------|-------|----------------------|----------|
| ry     | .760959   | .5179994  | 1.47  | 0.142 | -.2543011            | 1.776219 |
| rwp    | -1.682078 | .3724194  | -4.52 | 0.000 | -2.412006            | -.952149 |
| rap    | 2.958551  | .7114935  | 4.16  | 0.000 | 1.564049             | 4.353053 |
| rgp    | 2.299172  | .6349926  | 3.62  | 0.000 | 1.054609             | 3.543734 |
| _cons  | -3.212068 | 6.960658  | -0.46 | 0.644 | -16.85471            | 10.43057 |
| /sigma | .9260598  | .0686138  | 13.50 | 0.000 | .7915792             | 1.06054  |

$$\log(wg_t) = \beta_0 + \beta_1 \log(ry_t) + \beta_2 \log(rwp_t) + \beta_3 \log(rap_t) + \beta_4 \log(rgp_t)$$

Pick up the cases of  $wg_t > 10$ .