Description Remarks and examples Also see

## Description

In this example, we show how to estimate and interpret the results of an extended regression model with a binary outcome and endogenous sample selection.

## **Remarks and examples**

We are interested in whether regular exercise and body mass index (BMI) influence the chance of having a subsequent heart attack. In our fictional study, we collected data on 625 men who had a heart attack when they were between the ages of 50 and 55. Some men withdrew from the study before it completed, and we believe their reasons for leaving are related to unobserved factors that also affect their chances of having a second heart attack. We did, however, observe all cases where a second heart attack was fatal.

To account for the endogenous sample selection, we specify an auxiliary model for selection using a covariate that belongs in the auxiliary model and is excluded from the main equation. We expect that the direct effect of whether a man had regular checkups before the study is negligible after we condition on other covariates.

The outcome of interest is whether the man had another heart attack within five years of his first heart attack (attack). We believe that the man's current age is also an important exogenous covariate along with BMI. We model the indicator for whether the man was observed for the full five years of the study (full) as a function of an indicator for having regular checkups along with the covariates from the main equation.

```
. use https://www.stata-press.com/data/r19/heartsm
(Heart attacks)
. eprobit attack age bmi i.exercise, select(full = age bmi i.checkup) vce(robust)
Iteration 0: Log pseudolikelihood = -409.23137
              Log pseudolikelihood = -408.78569
Iteration 1:
Iteration 2:
              Log pseudolikelihood = -408.78452
Iteration 3: Log pseudolikelihood = -408.78452
Extended probit regression
                                                         Number of obs =
                                                                              625
                                                               Selected =
                                                                              458
                                                            Nonselected =
                                                                              167
                                                         Wald chi2(3)
                                                                        = 142.85
Log pseudolikelihood = -408.78452
                                                         Prob > chi2
                                                                        = 0.0000
                              Robust
               Coefficient
                             std. err.
                                                 P>|z|
                                                            [95% conf. interval]
                                            z
attack
                  .2237091
                                          6.37
                                                 0.000
                                                            .1548489
                                                                         .2925693
                             .0351334
         age
                  .1760896
                             .0298853
                                          5.89
                                                 0.000
                                                            .1175155
                                                                         .2346636
         bmi
    exercise
        Yes
                -1.438937
                             .1515198
                                         -9.50
                                                 0.000
                                                           -1.735911
                                                                       -1.141964
       _cons
                -15.78445
                             2.105945
                                         -7.50
                                                 0.000
                                                           -19.91202
                                                                       -11.65687
```

full

age

bmi

checkup Yes

corr(e.full, e.attack)

\_cons

-.1599347

-.1146582

2.306638

11.66488

-.4537026

.032953

.0208896

.1660248

1.942686

.1636665

We estimate that the correlation between the errors from the outcome equation and the errors from the selection equation is -0.45. This is significantly different from zero, so selection into the study is endogenous. Because the correlation is negative, we conclude that unobserved factors that increase the chance of staying in the study tend to occur with unobserved factors that decrease the chance of having a subsequent heart attack.

-4.85

-5.49

13.89

6.00

-2.77

0.000

0.000

0.000

0.000

0.006

-.2245214

-.1556011

1.981236

7.857284

-.71301

-.095348

-.0737152

2.632041

15.47247

-.0852183

The results for the main outcome equation (attack) and auxiliary selection equation (full) are interpreted just as you would those from hcckprobit. Which is to also say that the results for the main equation can be interpreted as you would those from a probit regression using probit on uncensored data. The goal of including a selection model is to estimate the parameters of the main equation as though there were no selection.

Age and BMI have increased the chances of having another heart attack, while regular exercise decreases the chances. However, the magnitude of the effect on the probability of another heart attack cannot be determined from the coefficient estimates themselves. We can use margins to examine the effect of different covariates on the probability of having a second heart attack. But first we want to investigate a possible further complication in our data: regular exercise may be an endogenous treatment. We explore this in [ERM] Example 4b.

## Also see

- [ERM] eprobit Extended probit regression
- [ERM] eprobit postestimation Postestimation tools for eprobit and xteprobit
- [ERM] Intro 4 Endogenous sample-selection features
- [ERM] Intro 9 Conceptual introduction via worked example

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