

Notes for laboratory session 10

Logistic regression analysis

Use the contraceptive use dataset and repeat the logistic regression analysis of the two-factor (“more”, “age”) additive model:

```

. char more[omit] 0

. xi: logit cuse i.more i.age [freq=N], nolog
i.more           Imore_0-1      (naturally coded; Imore_0 omitted)
i.age            Iage_1-4       (naturally coded; Iage_1 omitted)

Logit estimates                                         Number of obs =      1607
                                                       LR chi2(4)    =     128.88
                                                       Prob > chi2   =     0.0000
                                                       Pseudo R2    =     0.0643

-----
          cuse |      Coef.    Std. Err.      z     P>|z|      [95% Conf. Interval]
-----+
  Imore_1 |   -.824092   .1171128   -7.037   0.000    -1.053629   -.5945552
  Iage_2 |   .3678306   .1753673    2.097   0.036     .024117   .7115443
  Iage_3 |   .8077888   .1597533    5.056   0.000     .494678   1.1209
  Iage_4 |   1.022618   .2039337    5.014   0.000     .6229158   1.422321
  _cons |  -.8698414   .1571298   -5.536   0.000    -1.17781   -.5618727
-----
```

Analysis as an ordinal logistic regression

Now use the ordinal logistic regression approach to fit the same model.

```

. xi: ologit cuse i.more i.age [freq=N], nolog table
i.more           Imore_0-1      (naturally coded; Imore_0 omitted)
i.age            Iage_1-4       (naturally coded; Iage_1 omitted)

Ordered logit estimates                                         Number of obs =      1607
                                                               LR chi2(4)    =     128.88
                                                               Prob > chi2   =     0.0000
                                                               Pseudo R2    =     0.0643

Log likelihood = -937.40449
-----
          cuse |      Coef.    Std. Err.      z     P>|z|      [95% Conf. Interval]
-----+
  Imore_1 |   -.824092   .1171128   -7.037   0.000    -1.053629   -.5945552
  Iage_2 |   .3678306   .1753673    2.097   0.036     .024117   .7115443
  Iage_3 |   .8077888   .1597533    5.056   0.000     .494678   1.1209
  Iage_4 |   1.022618   .2039337    5.014   0.000     .6229158   1.422321
-----+
  _cut1 |   .8698414   .1571298               (Ancillary parameter)
-----
```

cuse	Probability	Observed
No	$\Pr(xb+u < \text{cut1})$	0.6845
Yes	$\Pr(\text{cut1} < xb+u)$	0.3155

a) Compare the coefficients produced by the “logit” and the “ologit” STATA command.

Compare the results given by the table option in the logit command output with the following table.

```
. tab cuse [freq=N]
```

Contracepti ve use (Yes/No)	Freq.	Percent	Cum.
No	1100	68.45	68.45
Yes	507	31.55	100.00
Total	1607	100.00	

Predict the probabilities of contraceptive use and non-use via STATA predict command.

```
. predict p0 p1  
(option p assumed; predicted probabilities)
```

b) Use hand calculations in order to obtain the predicted probability of contraceptive use among women 25-29 years old desiring more children. Verify your result using the listing below.

```
. list age educat more cuse p0 p1 if age==2 & more==1
```

	age	educat	more	cuse	p0	p1
9.	25-29	High	Yes	Yes	.7901953	.2098047
10.	25-29	Low	Yes	No	.7901953	.2098047
11.	25-29	High	Yes	No	.7901953	.2098047
16.	25-29	Low	Yes	Yes	.7901953	.2098047

Probit regression analysis

Now perform the same analysis using probit regression

```
. xi: probit cuse i.age i.more [freq=N], nolog
i.age           Iage_1-4      (naturally coded; Iage_1 omitted)
i.more          Imore_0-1    (naturally coded; Imore_0 omitted)

Probit estimates                                         Number of obs     =      1607
                                                               LR chi2(4)      =     127.51
                                                               Prob > chi2     =     0.0000
                                                               Pseudo R2       =     0.0636

Log likelihood = -938.09112
```

cuse	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Iage_2	.2086109	.1003457	2.079	0.038	.0119369 .405285
Iage_3	.4685637	.0928326	5.047	0.000	.2866152 .6505122
Iage_4	.6048679	.1226446	4.932	0.000	.3644889 .8452469
Imore_1	-.4964618	.0714451	-6.949	0.000	-.6364916 -.3564319
_cons	-.515345	.0922618	-5.586	0.000	-.6961748 -.3345152

c) Compare the results of the probit and the logit model. Recall however that the logit coefficients are not standardized but must be divided $\pi/\sqrt{3}$.

Ordinal regression

Use the tumour data set and fit a bivariate (sex , therapy) ordinal regression model for the outcome variable “outc”.

```
. char sex[omit] 2

. xi: ologit outc i.sex i.therapy, nolog tab
i.sex           Isex_1-2      (naturally coded; Isex_2 omitted)
i.therapy       Ithera_0-1   (naturally coded; Ithera_0 omitted)

Ordered logit estimates                               Number of obs =      299
                                                       LR chi2(2)    =     10.91
                                                       Prob > chi2  =    0.0043
Log likelihood = -394.52832                         Pseudo R2   =    0.0136

-----
          outc |      Coef.    Std. Err.      z     P>|z|      [95% Conf. Interval]
-----+
Isex_1 |   .5413938   .2871816     1.885   0.059    -.0214717   1.104259
Ithera_1 |  -.580685   .2121478    -2.737   0.006    -.9964871  -.164883
-----+
_cut1 |  -.7766492   .2880856             (Ancillary parameters)
_cut2 |   .7906273   .2866223
_cut3 |   1.84145   .3056123
-----+

          outc |      Probability      Observed
-----+
Progress |   Pr(xb+u<_cut1)    0.2843
No chang |   Pr(_cut1<xb+u<_cut2) 0.3612
Partial |   Pr(_cut2<xb+u<_cut3) 0.1906
Complete |   Pr(_cut3<xb+u)    0.1639
```

d) What is the interpretation of the model coefficients?

Probit analysis

Analyze the previous example using probit analysis.

```
. xi:oprobit outc i.sex i.therapy, nolog
i.sex           Isex_1-2      (naturally coded; Isex_2 omitted)
i.therapy       Ithera_0-1   (naturally coded; Ithera_0 omitted)

Ordered probit estimates                               Number of obs =      299
                                                       LR chi2(2)    =     10.79
                                                       Prob > chi2  =    0.0045
Log likelihood = -394.5871                         Pseudo R2   =    0.0135

-----
          outc |      Coef.    Std. Err.      z     P>|z|      [95% Conf. Interval]
-----+
Isex_1 |   .3401406   .174902     1.945   0.052    -.002661   .6829422
Ithera_1 |  -.3344764   .125435    -2.667   0.008    -.5803245  -.0886282
-----+
_cut1 |  -.459358   .176613             (Ancillary parameters)
_cut2 |   .5050695   .1760197
_cut3 |   1.122025   .1836877
```

e) Compare the results of the previous two approaches.