

Sociology 7704: Regression Models for Categorical Data

Instructor: Natasha Sarkisian

Introduction to OLS Regression in Stata

To run an OLS regression:

```
. reg agekdbrn educ born sex mapres80
```

Source	SS	df	MS	Number of obs =	1091
Model	4954.03533	4	1238.50883	F(4, 1086) =	51.24
Residual	26251.1232	1086	24.172305	Prob > F =	0.0000
				R-squared =	0.1588
				Adj R-squared =	0.1557
				Root MSE =	4.9165
Total	31205.1586	1090	28.6285858		

agekdbrn	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
educ	.6122718	.0569422	10.75	0.000	.5005426	.724001
born	1.360161	.5816506	2.34	0.020	.2188749	2.501447
sex	-2.37973	.3075642	-7.74	0.000	-2.983218	-1.776243
mapres80	.0243138	.0119552	2.03	0.042	.0008558	.0477718
_cons	16.95808	1.101139	15.40	0.000	14.79748	19.11868

Note that regression coefficients are partial slope coefficients; they indicate the change in the expected value of the dependent variable associated with one unit increase in the independent variable, when all other independent variables are held constant. These coefficients can potentially have two types of interpretation: cross-sectional and over time. Strictly speaking, all analyses we will do in this course are based on cross-sectional data.

To interpret the results, let's see how born and sex are coded:

```
. codebook born sex
```

```
-----
born                                     was r born in this country
-----
      type:  numeric (byte)
      label:  born

      range:  [1,2]
unique values: 2                                units:  1
                                          missing .:  6/2765

      tabulation:  Freq.   Numeric   Label
                   2503         1    yes
                   256          2    no
                   6             .

-----
sex                                     respondents sex
-----
      type:  numeric (byte)
      label:  sex

      range:  [1,2]
unique values: 2                                units:  1
                                          missing .:  0/2765

      tabulation:  Freq.   Numeric   Label
                   1228         1    male
                   1537         2    female
```

To get standardized regression coefficients, we can use beta option:

```
. reg agekdbnrn educ born sex mapres80, beta
```

Source	SS	df	MS		Number of obs =	1091
Model	4954.03533	4	1238.50883		F(4, 1086) =	51.24
Residual	26251.1232	1086	24.172305		Prob > F =	0.0000
					R-squared =	0.1588
					Adj R-squared =	0.1557
Total	31205.1586	1090	28.6285858		Root MSE =	4.9165

agekdbnrn	Coef.	Std. Err.	t	P> t	Beta
educ	.6122718	.0569422	10.75	0.000	.3108984
born	1.360161	.5816506	2.34	0.020	.0651372
sex	-2.37973	.3075642	-7.74	0.000	-.2154051
mapres80	.0243138	.0119552	2.03	0.042	.0588174
_cons	16.95808	1.101139	15.40	0.000	.

These coefficients indicate the number of standard deviations that agekdbnrn increases per each one standard deviation increase in an independent variable.

Note on missing data: Stata estimation commands (e.g. regress, logit etc) automatically drop from the analysis all cases that miss data points on at least one of the variables used in the analyses (this is called listwise deletion). This can be very problematic when there is a lot of missing data and when the patterns of missing data are systematic (which is often the case); we will discuss the topic of dealing with missing data towards the end of the course.

In order to get your regression output to look nice, you can use estimates table. For example, for our regression model, we can run:

```
. est table,star b(%8.3f) varlabel stats(N) varwidth(40)
```

Variable	active
highest year of school completed	0.612***
was r born in this country	1.360*
respondents sex	-2.380***
mothers occupational prestige score (198	0.024*
Constant	16.958***
N	1091

legend: * p<0.05; ** p<0.01; *** p<0.001

This way you don't need to retype anything – it's closer to the journal format table. You can also have multiple models in the same table by storing the estimates of each and then using est table:

```
. quietly reg agekdbnrn educ born sex mapres80
. est store model1
. qui reg agekdbnrn educ born sex mapres80 sibs maeduc
. est store model2
. est table model1 model2, star b(%8.3f) varlabel stats(N) varwidth(40)
```

```

-----
Variable |      model1      model2
-----+-----
highest year of school completed |      0.612***      0.579***
was r born in this country |      1.360*      1.423*
respondents sex |     -2.380***     -2.381***
mothers occupational prestige score (198 |      0.024*      0.016
number of brothers and sisters |             -0.128*
highest year school completed, mother |             0.029
Constant |     16.958***     17.820***
-----+-----
N |          1091          1035
-----
legend: * p<0.05; ** p<0.01; *** p<0.001

```

To find out more details and options, see help est_table as well as help estout (that is another command that allows to format the output in more complex ways).

If you are using nominal variables with more than just 2 categories or ordinal independent variables, you should not enter these variables in the model the same way you would use a continuous variable. For a nominal variable, that will result in nonsensical coefficients, because the categories are not really placed in any order so one unit increase is meaningless. For an ordinal variable, it's a stretch to use it in that fashion, because we assume equal distances among all categories. Before assuming that, we should test that assumption by introducing categories as separate variables. Here's how that's done in Stata.

```

. codebook marital
-----
marital                                     marital status
-----+-----
type:  numeric (byte)
      label:  marital
      range:  [1,5]
unique values:  5
units:  1
missing .:  0/2765

      tabulation:  Freq.   Numeric   Label
                   1269         1   married
                   247         2   widowed
                   445         3   divorced
                   96         4   separated
                   708         5   never married

. reg agekdbrn educ born sex mapres80 i.marital
-----+-----
Source |      SS      df      MS
-----+-----
Model |  5991.99195      8  748.998994
Residual | 25213.1666  1082  23.3023721
-----+-----
Total | 31205.1586  1090  28.6285858

Number of obs =      1091
F( 8, 1082) =      32.14
Prob > F =      0.0000
R-squared =      0.1920
Adj R-squared =      0.1860
Root MSE =      4.8273

-----+-----
agekdbrn |      Coef.   Std. Err.      t    P>|t|     [95% Conf. Interval]
-----+-----
educ |   .5662673   .0570585     9.92   0.000    .4543094    .6782251
born |   1.317066   .5740325     2.29   0.022    .1907232    2.443409
sex |  -2.187909   .306421    -7.14   0.000   -2.789156   -1.586662
mapres80 | .0232956   .0117729     1.98   0.048    .0001953    .0463958
marital |

```

widowed		.331999	.5584542	0.59	0.552	-.7637768	1.427775
divorced		-.8996868	.3914891	-2.30	0.022	-1.667851	-.1315229
separated		-2.101723	.7018116	-2.99	0.003	-3.478789	-.7246572
never married		-2.76481	.4698441	-5.88	0.000	-3.686719	-1.842901
_cons		17.93003	1.111328	16.13	0.000	15.74943	20.11063

To change the reference group (omitted category):

```
. reg agekdbn educ born sex mapres80 ib5.marital
```

Source	SS	df	MS	Number of obs =	1091
Model	5991.99195	8	748.998994	F(8, 1082) =	32.14
Residual	25213.1666	1082	23.3023721	Prob > F =	0.0000
				R-squared =	0.1920
				Adj R-squared =	0.1860
Total	31205.1586	1090	28.6285858	Root MSE =	4.8273

agekdbn	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
educ	.5662673	.0570585	9.92	0.000	.4543094 .6782251
born	1.317066	.5740325	2.29	0.022	.1907232 2.443409
sex	-2.187909	.306421	-7.14	0.000	-2.789156 -1.586662
mapres80	.0232956	.0117729	1.98	0.048	.0001953 .0463958
marital					
married	2.76481	.4698441	5.88	0.000	1.842901 3.686719
widowed	3.096809	.6574738	4.71	0.000	1.806741 4.386877
divorced	1.865123	.5388517	3.46	0.001	.8078108 2.922436
separated	.663087	.7909822	0.84	0.402	-.8889456 2.21512
_cons	15.16522	1.133172	13.38	0.000	12.94176 17.38868

Alternatively, you can create permanent dichotomies for all categories:

```
. tab marital, gen(marital)
```

marital	status	Freq.	Percent	Cum.
	married	1,269	45.90	45.90
	widowed	247	8.93	54.83
	divorced	445	16.09	70.92
	separated	96	3.47	74.39
	never married	708	25.61	100.00
Total		2,765	100.00	

```
. des marital*
```

variable name	storage type	display format	value label	variable label
marital	byte	%8.0g	marital	marital status
marital1	byte	%8.0g		marital==married
marital2	byte	%8.0g		marital==widowed
marital3	byte	%8.0g		marital==divorced
marital4	byte	%8.0g		marital==separated
marital5	byte	%8.0g		marital==never married

```
. reg agekdbnr educ born sex mapres80 marital2 marital3 marital4 marital5
```

Source	SS	df	MS	Number of obs = 1091		
Model	5991.99195	8	748.998994	F(8, 1082)	=	32.14
Residual	25213.1666	1082	23.3023721	Prob > F	=	0.0000
				R-squared	=	0.1920
				Adj R-squared	=	0.1860
Total	31205.1586	1090	28.6285858	Root MSE	=	4.8273

agekdbnr	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
educ	.5662673	.0570585	9.92	0.000	.4543094	.6782251
born	1.317066	.5740325	2.29	0.022	.1907232	2.443409
sex	-2.187909	.306421	-7.14	0.000	-2.789156	-1.586662
mapres80	.0232956	.0117729	1.98	0.048	.0001953	.0463958
marital2	.331999	.5584542	0.59	0.552	-.7637768	1.427775
marital3	-.8996868	.3914891	-2.30	0.022	-1.667851	-.1315229
marital4	-2.101723	.7018116	-2.99	0.003	-3.478789	-.7246572
marital5	-2.76481	.4698441	-5.88	0.000	-3.686719	-1.842901
_cons	17.93003	1.111328	16.13	0.000	15.74943	20.11063

Hypothesis testing after regression models

Oftentimes, we need to test hypotheses to find out whether coefficients are jointly significant, or to assess whether certain coefficients are equal to each other.

```
. reg agekdbnr educ born sex mapres80 i.marital
```

Source	SS	df	MS	Number of obs = 1091		
Model	5991.99195	8	748.998994	F(8, 1082)	=	32.14
Residual	25213.1666	1082	23.3023721	Prob > F	=	0.0000
				R-squared	=	0.1920
				Adj R-squared	=	0.1860
Total	31205.1586	1090	28.6285858	Root MSE	=	4.8273

agekdbnr	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
educ	.5662673	.0570585	9.92	0.000	.4543094	.6782251
born	1.317066	.5740325	2.29	0.022	.1907232	2.443409
sex	-2.187909	.306421	-7.14	0.000	-2.789156	-1.586662
mapres80	.0232956	.0117729	1.98	0.048	.0001953	.0463958
marital						
widowed	.331999	.5584542	0.59	0.552	-.7637768	1.427775
divorced	-.8996868	.3914891	-2.30	0.022	-1.667851	-.1315229
separated	-2.101723	.7018116	-2.99	0.003	-3.478789	-.7246572
never married	-2.76481	.4698441	-5.88	0.000	-3.686719	-1.842901
_cons	17.93003	1.111328	16.13	0.000	15.74943	20.11063

```
. ereturn list
```

scalars:

```

      e(N) = 1091
    e(df_m) = 8
    e(df_r) = 1082
      e(F) = 32.14260719876214
      e(r2) = .1920192757413473
    e(rmse) = 4.827253060400016
    e(mss) = 5991.991948028161

```

```

        e(rss) = 25213.16662209091
        e(r2_a) = .1860452962643886
        e(ll) = -3261.080117285556
        e(ll_0) = -3377.390032713303
        e(rank) = 9

macros:
        e(cmdline) : "regress agekdbrn educ born sex mapres80 i.marital"
        e(title) : "Linear regression"
        e(marginsok) : "XB default"
        e(vce) : "ols"
        e(depvar) : "agekdbrn"
        e(cmd) : "regress"
        e(properties) : "b V"
        e(predict) : "regres_p"
        e(model) : "ols"
        e(estat_cmd) : "regress_estat"

matrices:
        e(b) : 1 x 10
        e(V) : 10 x 10

functions:
        e(sample)

. mat list e(b)

e(b) [1,10]
      educ      born      sex      mapres80      1b.      2.      3.      4.      5.
yl      .56626729      1.3170661      -2.1879089      .02329556      marital      marital      marital      marital      marital      _cons
      0      .33199897      -.89968677      -2.1017231      -2.7648101      17.930029

```

A joint significance test:

```

. test 2.marital 3.marital 4.marital 5.marital

( 1)  2.marital = 0
( 2)  3.marital = 0
( 3)  4.marital = 0
( 4)  5.marital = 0

      F( 4, 1082) = 11.14
      Prob > F = 0.0000

```

Testing whether categories 2, 3, and 4 can be combined:

```

. test 2.marital=3.marital

( 1)  2.marital - 3.marital = 0

      F( 1, 1082) = 4.00
      Prob > F = 0.0457

. test 2.marital=4.marital, acc

( 1)  2.marital - 3.marital = 0
( 2)  2.marital - 4.marital = 0

      F( 2, 1082) = 4.34
      Prob > F = 0.0133

```

For an ordinal variable, hypothesis testing allows us to evaluate whether each one unit increase produces the same change in the dependent variable:

. reg agekdbrn educ born sex mapres80 i.degree				Number of obs = 1091	
Source	SS	df	MS	F(8, 1082)	= 32.94
Model	6111.91384	8	763.98923	Prob > F	= 0.0000
Residual	25093.2447	1082	23.1915386	R-squared	= 0.1959
Total	31205.1586	1090	28.6285858	Adj R-squared	= 0.1899
				Root MSE	= 4.8158

agekdbrn	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
educ	.0506574	.1089486	0.46	0.642	-.163117	.2644317
born	1.267439	.570358	2.22	0.026	.1483064	2.386572
sex	-2.192157	.3025278	-7.25	0.000	-2.785764	-1.598549
mapres80	.0225168	.0118318	1.90	0.057	-.0006991	.0457326
degree						
high school	1.934153	.6048514	3.20	0.001	.7473387	3.120968
junior college	2.201938	.8713455	2.53	0.012	.4922196	3.911656
bachelor	4.446438	.9701565	4.58	0.000	2.542837	6.350039
graduate	7.624749	1.215111	6.27	0.000	5.240509	10.00899
_cons	21.78773	1.329524	16.39	0.000	19.17899	24.39647

The increases are 1.93, 0.27, 2.24, 3.18, so they look unequal (especially because the junior college category is not really different from high school category), but let's test it:

```
. test 1.degree=(2.degree-1.degree)

( 1)  2*1.degree - 2.degree = 0

      F( 1, 1082) =    4.53
      Prob > F =    0.0335

. test 1.degree=(3.degree-2.degree), acc

( 1)  2*1.degree - 2.degree = 0
( 2)  1.degree + 2.degree - 3.degree = 0

      F( 2, 1082) =    2.41
      Prob > F =    0.0905

. test 1.degree=(4.degree-3.degree), acc

( 1)  2*1.degree - 2.degree = 0
( 2)  1.degree + 2.degree - 3.degree = 0
( 3)  1.degree + 3.degree - 4.degree = 0

      F( 3, 1082) =    3.98
      Prob > F =    0.0078
```