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Stata tip 58: nl is not just for nonlinear models

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1 Introduction

The nl command makes performing nonlinear least-squares estimation almost as easy as performing linear regression. In this tip, three examples are given where nl is preferable to regress, even when the model is linear in the parameters.

2 Transforming independent variables

Using the venerable auto dataset, suppose we want to predict the weight of a car based on its fuel economy measured in miles per gallon. We first plot the data:

- . sysuse auto . scatter weight mpg
- Clearly, there is a negative relationship between weight and mpg, but is that relationship linear? The engineer in each of us believes that the amount of gasoline used to go one mile should be a better predictor of weight than the number of miles a car can go on one gallon of gas, so we should focus on the reciprocal of mpg. One way to proceed would be to create a new variable, gpm, measuring gallons of gasoline per mile and then to use regress to fit a model of weight on gpm. However, consider using nl instead:

```
. nl (weight = {b0} + {b1}/mpg)
(obs = 74)
Iteration 0: residual SS = 1.19e+07
Iteration 1:
              residual SS = 1.19e+07
                     SS
      Source
                               df
                                        MS
       Model
                32190898.6
                                1 32190898.6
                                                       R-squared
                                                                           0.7300
    Residual
                11903279.8
                                   165323.33
                                                       Adj R-squared =
                                                                           0.7263
                                                       Root MSE
                                                                     = 406.5997
       Total
                44094178.4
                               73 604029.841
                                                       Res. dev.
                                                                        1097.134
                                                  P>|t|
                                                            [95% Conf. Interval]
                            Std. Err.
      weight
                    Coef.
                                            t
         /b0
                 415.1925
                             192.5243
                                          2.16
                                                  0.034
                                                            31.40241
                                                                        798.9826
         /b1
                 51885.27
                             3718.301
                                         13.95
                                                  0.000
                                                            44472.97
                                                                        59297.56
```

Parameter b0 taken as constant term in model & ANOVA table

140 Stata tip 58

(You can verify that R^2 from this model is higher than that from a linear model of weight on mpg. You can also verify that our results match those from regressing weight on gpm.)

Here a key advantage of nl is that we do not need to create a new variable containing the reciprocal of mpg. When doing exploratory data analysis, we might want to consider using the natural log or square root of a variable as a regressor, and using nl saves us some typing in these cases. In general, instead of typing

```
. generate sqrtx = sqrt(x)
. regress y sqrtx
we can type
. nl (y = {b0} + {b1}*sqrt(x))
```

3 Marginal effects and elasticities

Using n1 has other advantages as well. In many applications, we include not just the variable x in our model but also x^2 . For example, most wage equations express log wages as a function of experience and experience squared. Say we want to fit the model

$$y_i = \alpha + \beta_1 x_i + \beta_2 x_i^2 + \epsilon_i$$

and then determine the elasticity of y with respect to x; that is, we want to know the percent by which y will change if x changes by one percent.

Given the interest in an elasticity, the inclination might be to use the mfx command with the eyex option. We might type

```
. generate xsq = x^2
. regress y x xsq
. mfx compute, eyex
```

These commands will not give us the answer we expect because regress and mfx have no way of knowing that xsq is the square of x. Those commands just see two independent variables, and mfx will return two "elasticities", one for x and one for xsq. If x changes by some amount, then clearly x^2 will change as well; however, mfx, when computing the derivative of the regression function with respect to x, holds xsq fixed!

The easiest way to proceed is to use nl instead of regress:

```
. nl (y = {a} + {b1}*x + {b2}*x^2), variables(x) 
 . mfx compute, eyex
```

Whenever you intend to use mfx after nl, you must use the variables() option. This option causes nl to save those variable names among its estimation results.

B. P. Poi 141

4 Constraints

nl makes imposing nonlinear constraints easy. Say you have the linear regression model

$$y_i = \alpha + \beta_1 x_{1i} + \beta_2 x_{2i} + \beta_3 x_{3i} + \epsilon_i$$

and for whatever reason you want to impose the constraint that $\beta_2\beta_3 = 5$. We cannot use the constraint command in conjunction with regress because constraint only works with linear constraints. n1, however, provides an easy way out. Our constraint implies that $\beta_3 = 5/\beta_2$, so we can type

. nl (y = {a} + {b1}*x1 + {b2=1}*x2 +
$$(5/{b2})*x3$$
)

Here we initialized β_2 to be 1 because if the product of β_2 and β_3 is not 0, then neither of those parameters can be 0, which is the default initial value used by nl.