

Stata

Stata is a statistical software with two notable features:

- It can be used interactively through its graphical interface and command input box, or run pre-written scripts. This makes it easier to learn than many other statistical software.
- It only operates on one dataset at a time and all data is loaded into memory. This allows Stata to operate faster than harddisk-based software such as SAS, but you can run into space problem if your dataset is very, very large.

The screenshot shows the Stata MP 13.1 interface with several callout boxes:

- Viewer shows you Stata's help files.** (Points to the top-left toolbar area)
- Do-File Editor is a built-in text editor to edit command files.** (Points to the top-left toolbar area)
- Data Editor & Browser shows you the data currently in Stata's memory.** (Points to the top-left toolbar area)
- Variables lists all your variables.** (Points to the Variables list on the right)
- Review lists out commands that have previously been executed. Successful executions are in black while failed ones are in red. Double-click on any of them to re-execute. You can also use the [Page-up] and [page-down] keys to move through the list.** (Points to the Review window on the left)
- Results shows you Stata's display output.** (Points to the main Results window)
- Command is where you enter commands.** (Points to the Command input box at the bottom)
- Properties lists the properties of the currently-selected variable.** (Points to the Properties window on the right)

1. Log File: Keeping a record of the commands used and the results generated.

| Description | Command | Example |
|--|-----------------------------------|---------------------------------------|
| Start a log: record your session into a file called a log file | log using "filename", text | log using "D:\Economics\log.txt",text |
| Close log | log close | log close |

2. Importing: We can import excel files into Stata's Data Editor and then save them as dta format.

| Description | Command | Example |
|---|--|--|
| Import an Excel file, also known as a workbook, into Stata's Data Editor | import excel using "filename" | import excel using "D:\Economics\company_record.xlsx" |
| Import an Excel file and treat the first row as variable names | import excel using "filename", firstrow | import excel using "D:\Economics\company_record.xlsx", firstrow |
| Save the workbook into a dta format | save "filename" | save "D:\Economics\company_record" |
| Import another excel file into Stata's Data Editor Note: Data editor cannot contain two datasets, so we need to clear the previous one | Import excel using "filename", firstrow clear | import excel using "D:\Economics\employee_survey.xlsx", firstrow clear |
| Save the workbook into a dta format | save "filename" | save "D:\Economics\employee_survey" |
| If the dta file already exists, overwrite with <i>replace</i> | save "filename", replace | save "D:\Economics\employee_survey", replace |

3a. Use/ load a Stata dataset (dta format)

| Description | Command | Example |
|--|-------------------------------------|--|
| First, clear the data in the data editor | clear | clear |
| load a dta file in the data editor | use "filename" | use "D:\Economics\company_record" |
| The above two steps can be combined into one command | use "filename", clear | use "D:\Economics\company_record", clear |

Note: A newer version of Stata can open datasets saved by an older version of Stata, but the reverse is not true.

3b. Change Working Directory

| Description | Command | Example |
|--|-----------------------|----------------------|
| Alternatively, we can first change the working directory before loading a stata dataset, to avoid typing again the full address. | cd "directory" | cd "D:\Economics\" |
| Then, load a dta file in the data editor | use "filename" | use "company_record" |

3c. Merging Datasets

| Description | Command | Example |
|---|---|--------------------------------------|
| Merge: Merging a dataset to another dataset in the memory of the Data Editor, matching on one or more key variables | merge 1:1 variables using "filename" merge 1:m variables using "filename" merge m:1 variables using "filename" merge m:m variables using "filename" | merge 1:1 id using "employee_survey" |
| Append: Adding data to bottom of the existing dataset | append using "filename" | append using "company_record_2" |

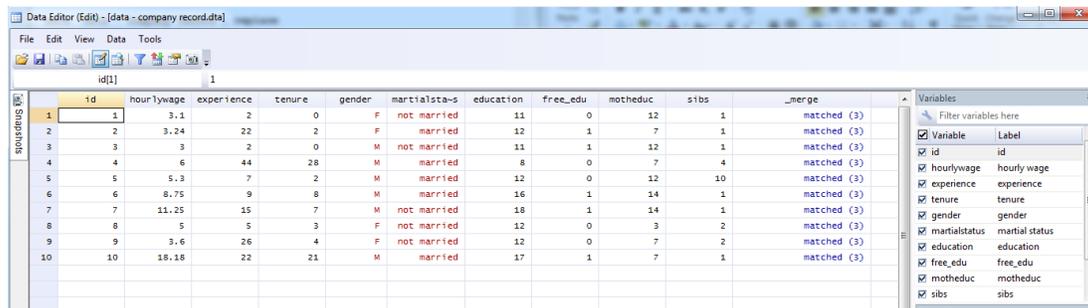


Fig. 2 Merged dataset

3d. Data frames (Stata 16 onwards)

| Description | Command | Example |
|-------------------------------|---|---|
| Create a frame | frame create name | frame create survey |
| Change current frame | frame change name | frame change survey frame change default |
| Delete frame | frame drop name | frame drop survey |
| Do something on a frame | frame name: ... frame name { ... } | frame survey: use employee_survey |
| Link with another frame | frlink m:n variables, frame(name) | frlink 1:1 id, frame(survey) |
| Fetch data from another frame | frget varname, from(name) | frget education, from(survey) |

4. Manipulate Data

| Description | Command | Example |
|--|--------------------------------------|--|
| Adding a new variable | generate <i>new_var</i> = ... | gen log_edu = log(education) |
| Modifying a variable | replace <i>variable</i> = ... | replace log_edu = ln(education) |
| Drop a variable | drop <i>variable</i> | drop log_edu |
| Drop an observation | drop if <i>variable</i> = ... | drop if id == 11 |
| Switch between the two common ways of storing groups of data | reshape ... | (Read Stata's help file if you need this function) |

Note: Type "function" in the viewer for a list of available functions.
 Stata follows the common programming convention of using "=" for assignment (i.e. modification of data) and "==" for comparison.

5. Summarize: to obtain summary statistics

| Description | Command | Example |
|--------------------------------|-------------------------------------|------------------------|
| Summarize | sum | sum |
| Summarize a variable | sum <i>variable</i> | sum hourlywage |
| Summarise a variable in detail | sum <i>variable</i> , detail | sum hourlywage, detail |

6. Making a table

| Description | Command | Example |
|--|--|--|
| Making a table of summary statistics: Make a table with certain contents | table <i>variable1 variable2</i> , contents(option) | table gender free_edu, contents(median hourlywage) table gender free_edu, contents(median hourlywage sd hourlywage) |

```

-----
      gender |      free_edu
             |      0      1
-----+-----
           F |   3.6   3.24
             |   3       1
           M |   5.65   10
             |   2       4
-----
    
```

An example of *Table* command output

7. Correlation

| Description | Command | Example |
|---|--|---|
| Correlations (covariances) of variables | correlate <i>variable1</i> <i>variable2 variable3...</i> | corr hourlywage experience education |

```

-----+-----
          | hourly~e experi~e educat~n
-----+-----
hourlywage | 1.0000
experience  | 0.1940  1.0000
education   | 0.7592 -0.2218  1.0000
    
```

An example of *Correlation* command output

8. T-Test

| Description | Command | Example |
|--|---|----------------------------|
| T-test: compare the means of two variables | ttest <i>variable1 = variable2</i> | ttest education = motheduc |

```

. ttest education = motheduc

Paired t test
-----+-----
Variable |   Obs   Mean   Std. Err.   Std. Dev.   [95% Conf. Interval]
-----+-----
educat~n |     10    12.9   .9826269    3.107339    10.67714    15.12286
motheduc |     10     9.5    1.185561    3.749074     6.818074    12.18193
-----+-----
diff |     10     3.4    1.240072    3.921451     .594763    6.205237
-----+-----
      mean(diff) = mean(education - motheduc)          t = 2.7418
Ho: mean(diff) = 0                                degrees of freedom = 9

Ha: mean(diff) < 0          Ha: mean(diff) != 0          Ha: mean(diff) > 0
Pr(T < t) = 0.9886          Pr(|T| > |t|) = 0.0228          Pr(T > t) = 0.0114
    
```

An example of *ttest* command output for test of two variables

| Description | Command | Example |
|--|---|------------------------------|
| T-test: compare the means of two groups within the same variable | ttest <i>variable1, by(groupvar)</i> | ttest hourlywage, by(gender) |

Note: *groupvar* can only take on two values

```

. ttest hourlywage, by(gender)

Two-sample t test with equal variances
-----+-----
Group |   Obs   Mean   Std. Err.   Std. Dev.   [95% Conf. Interval]
-----+-----
F |     4    3.735   .4346167    .8692334    2.351856    5.118144
M |     6    8.746667  2.218876    5.435114    3.042864    14.45047
-----+-----
combined |     10    6.742   1.528432    4.833326    3.284447    10.19955
-----+-----
diff |          -5.011667  2.794796          -11.45648    1.433145
-----+-----
      diff = mean(F) - mean(M)          t = -1.7932
Ho: diff = 0                                degrees of freedom = 8

Ha: diff < 0          Ha: diff != 0          Ha: diff > 0
Pr(T < t) = 0.0553          Pr(|T| > |t|) = 0.1107          Pr(T > t) = 0.9447
    
```

An example of *ttest* command output for test of two groups within the same variable

9a. Histogram

| Description | Command | Example |
|--|---|---------------------------------|
| Histogram of a variable | hist <i>variable</i> | hist education |
| Histogram of a variable, with n blocks (Fig 3) | hist <i>variable</i> , bin(n) | hist education, bin(5) |
| Histogram of a variable, with n blocks, and y axis as fraction (Fig 4) | hist <i>variable</i> , bin(n) fraction | hist education, bin(5) fraction |

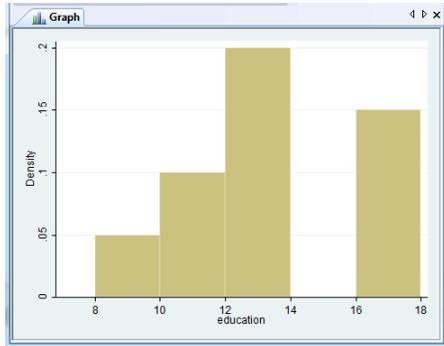


Fig 3

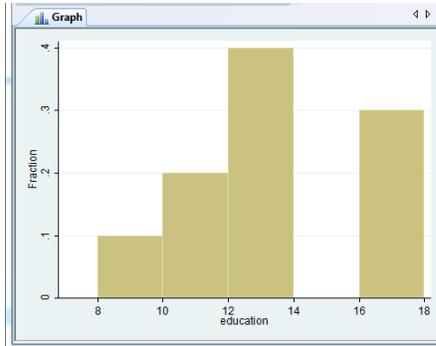


Fig 4

9b. Scatter Graph

| Description | Command | Example |
|--|--|--|
| Plot a scatter graph | scatter <i>variable1 variable2</i> | scatter hourlywage education |
| Plot two scatter subgraphes , being placed beside each other (Fig 5) | scatter <i>variable1 variable2</i> , by(variable3) | scatter hourlywage educ, by(gender) |
| Plot two subgraphes, one placing on another (Fig 6) | scatter <i>variable1 variable2</i> if <i>variable3</i> == value1 scatter <i>variable1 variable2</i> if <i>variable3</i> == value2 | scatter hourlywage educ if gender == "M" scatter hourlywage educ if gender == "F" |

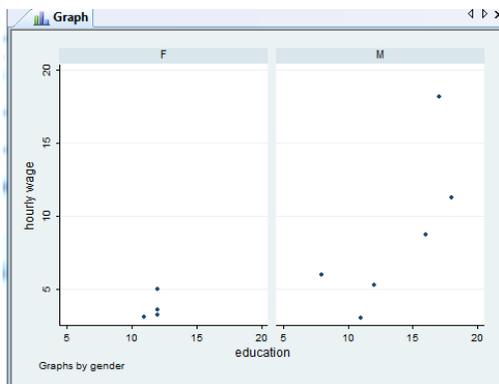


Fig 5

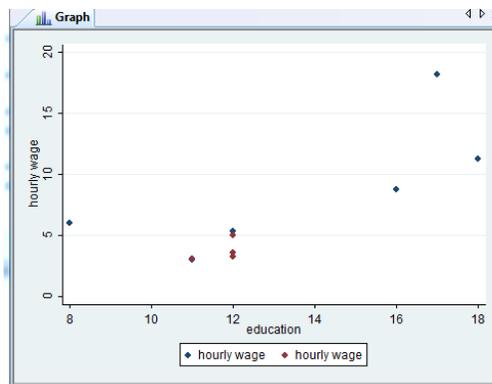


Fig 6

10a. Regression

| Description | Command | Example |
|-----------------------|--|-------------------------------------|
| Ordinary Least Square | regress <i>dep_variable</i> <i>indep_variables</i> | reg hourlywage experience tenure |

```
. reg hourlywage experience tenure
```

| Source | SS | df | MS | | | |
|----------|------------|----|------------|-----------------|--------|--|
| Model | 108.737034 | 2 | 54.3685168 | Number of obs = | 10 | |
| Residual | 101.512326 | 7 | 14.5017609 | F(2, 7) = | 3.75 | |
| Total | 210.24936 | 9 | 23.36104 | Prob > F = | 0.0782 | |
| | | | | R-squared = | 0.5172 | |
| | | | | Adj R-squared = | 0.3792 | |
| | | | | Root MSE = | 3.8081 | |

| hourlywage | Coef. | Std. Err. | t | P> t | [95% Conf. Interval] | |
|------------|-----------|-----------|-------|-------|----------------------|----------|
| experience | -.2491599 | .1541775 | -1.62 | 0.150 | -.6137318 | .115412 |
| tenure | .5712551 | .2166466 | 2.64 | 0.034 | .0589673 | 1.083543 |
| _cons | 6.294649 | 1.913857 | 3.29 | 0.013 | 1.769097 | 10.8202 |

An example of *regress* command

10b. Regression with dummy variables:

First, we have to generate dummy variables for qualitative variables

Step1: generate the name of dummy variable = 0 (Fig 7)
 generate gender_dummy = 0

Step2: replace the name of dummy variable = 1 if variable == "value1"
 replace gender_dummy = 1 if gender=="M"
 == "value1"
 replace the name of dummy variable = 2 if variable == "value2"
 == "value2" (Fig 8)

Alternatively, use **xi** xi i.gender

Then, we do regression with the dummy variables
 reg *variable1 variable2 the name of dummy variable*
 reg hourlywage experience tenure gender_dummy

| | motheduc | sibs | _merge | genderdummy |
|----|----------|------|-------------|-------------|
| 1 | 12 | 1 | matched (3) | 0 |
| 2 | 7 | 1 | matched (3) | 0 |
| 3 | 12 | 1 | matched (3) | 0 |
| 4 | 7 | 4 | matched (3) | 0 |
| 5 | 12 | 10 | matched (3) | 0 |
| 6 | 14 | 1 | matched (3) | 0 |
| 7 | 14 | 1 | matched (3) | 0 |
| 8 | 3 | 2 | matched (3) | 0 |
| 9 | 7 | 2 | matched (3) | 0 |
| 10 | 7 | 1 | matched (3) | 0 |

Fig 7

| | motheduc | sibs | _merge | genderdummy |
|----|----------|------|-------------|-------------|
| 1 | 12 | 1 | matched (3) | 0 |
| 2 | 7 | 1 | matched (3) | 0 |
| 3 | 12 | 1 | matched (3) | 1 |
| 4 | 7 | 4 | matched (3) | 1 |
| 5 | 12 | 10 | matched (3) | 1 |
| 6 | 14 | 1 | matched (3) | 1 |
| 7 | 14 | 1 | matched (3) | 1 |
| 8 | 3 | 2 | matched (3) | 0 |
| 9 | 7 | 2 | matched (3) | 0 |
| 10 | 7 | 1 | matched (3) | 1 |

Fig 8

11. Fixed-Effect Regression

| Description | Command | Example |
|--|--|--|
| If there are too many values for the dummy variable, we can encode the variable into numeric first | encode <i>variable</i> , generate(<i>new numeric dummy variable</i>) | encode martialstatus, generate(martialstatus_numeric) |
| Then, run fixed effect regression | xtreg <i>dep_variable indep_variables</i> , fe i(<i>new numeric dummy variable</i>) | xtreg hourlywage tenure, fe i(martialstatus_numeric) |

| | motheduc | sibs | _merge | genderdummy | martialsta~c |
|----|----------|------|-------------|-------------|--------------|
| 1 | 12 | 1 | matched (3) | 0 | not married |
| 2 | 7 | 1 | matched (3) | 0 | married |
| 3 | 12 | 1 | matched (3) | 1 | not married |
| 4 | 7 | 4 | matched (3) | 1 | married |
| 5 | 12 | 10 | matched (3) | 1 | married |
| 6 | 14 | 1 | matched (3) | 1 | married |
| 7 | 14 | 1 | matched (3) | 1 | not married |
| 8 | 3 | 2 | matched (3) | 0 | not married |
| 9 | 7 | 2 | matched (3) | 0 | not married |
| 10 | 7 | 1 | matched (3) | 1 | married |

Output of *encode*. The leftmost variable is in fact numeric, but is labeled.

```
. xtreg hourlywage tenure, fe i(martialstatus_numeric)

Fixed-effects (within) regression              Number of obs   =       10
Group variable: martialsta~c                  Number of groups =         2

R-sq:  within = 0.2532                        Obs per group:  min =         5
          between = 1.0000                      avg           =       5.0
          overall = 0.3370                      max           =         5

corr(u_i, Xb) = 0.5227                        F(1, 7)         =         2.37
                                          Prob > F        =       0.1673

-----+-----
hourlywage |      Coef.   Std. Err.      t    P>|t|     [95% Conf. Interval]
-----+-----
    tenure |   .2832131   .1838511     1.54   0.167    - .1515256   .7179517
    _cons |   4.617902   1.971669     2.34   0.052    - .0443534   9.280157
-----+-----
sigma_u |   .31239775
sigma_e |   4.4566344
rho     |   .0048896   (fraction of variance due to u_i)

F test that all u_i=0:   F(1, 7) =      0.02                Prob > F = 0.8975
```

An example of *xtreg* command

12a. Correction for Heteroskedasticity

| Description | Command | Example |
|--|--|--|
| Test if the homoscedasticity assumption holds (Run after <i>regress</i>) | estat <i>hettest indep_variables</i> | hettest experience tenure |
| <u>Robust Standard Errors</u> (Eicker-White Std. Err.) | <i>regress dep_variable indep_variables, robust</i> (Also works with <i>xtreg</i>) | reg hourlywage experience tenure, robust |

```
. reg hourlywage experience tenure, robust
```

```
Linear regression                               Number of obs =      10
                                                F( 2,      7) =      3.25
                                                Prob > F          =    0.1002
                                                R-squared         =    0.5172
                                                Root MSE         =    3.8081
```

```
-----+-----
```

| hourlywage | Coef. | Robust Std. Err. | t | P> t | [95% Conf. Interval] |
|------------|-----------|------------------|-------|-------|----------------------|
| experience | -.2491599 | .1273134 | -1.96 | 0.091 | -.5502081 .0518884 |
| tenure | .5712551 | .2351817 | 2.43 | 0.045 | .0151388 1.127371 |
| _cons | 6.294649 | 1.705796 | 3.69 | 0.008 | 2.261082 10.32822 |

```
-----+-----
```

An example of robust standard errors. Note the difference in standard errors compared to 10a.

12b. Correction for Error Correlation within Group and Over Time

| | | |
|---|---|--|
| <u>Clustered Standard Errors</u> Corrects within-group error correlation | <i>regress dep_variable indep_variables, vce(cluster clustervar)</i> (Also works with <i>xtreg</i>) | reg hourlywage tenure, vce(cluster workplace) |
| <u>Newey-West Standard Errors</u> Corrects for equi-correlated error over time. Error beyond the number of <i>periods</i> specified are assumed to be uncorrelated | <i>newey dep_variable indep_variables, lag(periods)</i> <u>Let Stata select optimal lag:</u> <i>ivregress gmm dep_var indep_vars, wmatrix(hac nw opt)</i> | newey hourlywage tenure, lag(2) ivregress gmm hourlywage tenure, wmat(hac nw opt) |

13. Hypothesis Testing

| | | |
|----------------------------|--|--|
| Test linear hypothesis | test <i>varnames</i> <i>test exp1 [= exp2 = ...]</i> | test tenure experience test tenure – experience = 0 |
| Test non-linear hypothesis | testnl <i>exp2 [= exp2 = ...]</i> | testnl _b[tenure]^2 = 0 |

14. Obtaining residuals and predicted values

| | | |
|--|--|------------------------------|
| Obtain predicted values after regression | predict <i>new_var</i> | predict predicted_hourlywage |
| Obtain residuals | predict <i>new_var, residuals</i> | predict estimated_u, r |

15. Instrumental Variable Regression

| Description | Command | Example |
|--|---|---|
| When an independent variable is correlated with the error term, OLS is biased. IV regression uses another variable uncorrelated with the error to predict the correlated one | ivregress estimator <i>dep_var</i> <i>exog_vars</i> (<i>endo_var</i> = <i>instrument_vars</i>) | ivregress 2sls hourlywage (education = free_educ) |
| Test for endogeneity after IV regression | estat endogenous | Estat endog |

16a. Discrete Choice Model

| Description | Command | Example |
|--|--|------------------------------|
| <u>Logit:</u> When the dependent variable takes on binary values, we can use the logit model | logit <i>dep_var indep_vars</i> | logit free_educ motheduc |
| However, the interpretation of β Estimator is different from the one we used for OLS. So we need to use odd ratios | logit <i>dep_var indep_vars</i> , or | logit free_educ motheduc, or |

16b. Additional Discrete Choice Models

| Description | Command | Example |
|---|--|--|
| <u>Multinomial Logit:</u> When the dependent variable takes on more than two discrete values | mlogit <i>dep_var indep_vars</i> | mlogit free_educ motheduc |
| <u>Ordered Logit:</u> When the dependent variable represents ordinal ratings (e.g. bad, good, best) | ologit <i>dep_var indep_vars</i> | ologit feedback budget, or |
| <u>Rank-ordered Logit:</u> When the dependent variable represents successive draws without replacement (e.g. places in a race) | rologit <i>dep_var indep_vars</i> , group(<i>horse_id</i>) | rologit position training, group(<i>horse_id</i>) |

17. Obtaining Marginal Effects

| Description | Command | Example |
|--|--|--|
| The marginal effect of each independent variable on the predicted value at the average value of the variable | <u>old syntax:</u> mf <u>new syntax:</u> margins , dydx(<i>indep_vars</i>) atmeans | mf margins, dydx(motheduc) atmeans |