

ECON 20: Lecture #3

Stata Introduction

1 Getting started

Your first step is to download the dataset `cpsmar08_10pct.dta` using the link on canvas. Create a new, empty folder on your computer called `stata_intro` and move the file you just downloaded into that folder. Take note of this new folder's "location" on your computer.

Once you have done this, you should launch STATA. The first thing we need to get comfortable with is the idea of a *working directory*. This is the location (i.e., folder) on your computer where STATA will work from. You should tell STATA to change the working directory to your new `stata_intro` folder using the `cd` command:

```
/// EXAMPLE ON MAC ///
```

```
cd "~/Dropbox/Econ20/stata_intro/"
```

```
/// EXAMPLE ON PC ///
```

```
cd "C:\Users\steve\Dropbox\Econ20\stata_intro\"
```

You will need to replace the directory locations after the `cd` to the folder names corresponding to the location of the folder on your own computer.

2 Loading the data

Now that we are working from the correct directory, we can load our dataset into STATA by typing the following into the command line:

```
use cpsmar08_10pct.dta, clear
```

Note: You can always open a dataset by double clicking on the file on your computer. On most systems, this will automatically change the working directory to the location of that file on your computer. You can also open a dataset by going to **File->Open** in STATA, but

I recommend against this practice. You need to get used to the idea of a working directory and doing *everything* by issuing commands in the command line.

Now is a good time to familiarize ourselves with the STATA environment. This will be discussed using a screenshare during class. The two most important panels in the STATA environment are the *Results* and *Command* panels. The command panel is where you type commands in for STATA to perform. The results panel displays the output corresponding to a given command.

3 Exploring the data

With a dataset loaded into STATA, you can view the dataset with the **browse** command. This opens up a spreadsheet-like view of the dataset. Some questions that we might have (or be able to answer) after using the **browse** command:

1. What kind of data are these?
2. Why are the words **male** and **female** colored blue? (*Hint*: Look at the codebook)
3. What do you think it means that some of the values of **wage_hr** seem to be a period?

Some other useful ways to get some basic facts about the dataset are the **describe** and **summarize** commands. The **describe** command (or **desc** for short) lists all the variables in the dataset, with information on storage type as well as value and variable labels. The **summarize** (or **sum** for short) provides basic summary statistics for all the variables in the dataset. Try entering the **desc** and **sum** commands into the command line and making sense of the output.

4 Basic operations

Creating variables

You can create new variables using the **generate** (or **gen** for short) command. Start by generate a new variable for the natural log of the wage:

```
gen lnwage = ln(wage_hr)
```

You can see the mean and standard deviation of your new variable (or any variable) in the dataset by issuing the **sum** command followed by the variable name. So if we issue the command **sum lnwage**, we get the output:

```
. summ lnwage
```

Variable	Obs	Mean	Std. Dev.	Min	Max
lnwage	1661	2.849478	.6470944	.2766323	5.441418

Regressions

You can estimate a regression using the `reg` command. The syntax for the regression command is `reg Y X`, where Y is the left-hand side variable and X is the right-hand side variable. Note that for multiple regression, we would use `reg Y X1 X2 X3`.

Start by estimating a regression of log wages on years of education using the command `reg lnwage yrsed`. This gives us the output:

```
. reg lnwage yrsed
```

Source	SS	df	MS	Number of obs = 1661		
Model	177.419565	1	177.419565	F(1, 1659) = 568.58		
Residual	517.674217	1659	.312039914	Prob > F = 0.0000		
				R-squared = 0.2552		
				Adj R-squared = 0.2548		
Total	695.093782	1660	.418731194	Root MSE = .55861		

lnwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
yrsed	.1177832	.0049396	23.84	0.000	.1080948	.1274717
_cons	1.214163	.0699375	17.36	0.000	1.076988	1.351338

Notice the key elements. The intercept ($\hat{\beta}_0$) is 1.214163. It has standard error .0699375. The slope ($\hat{\beta}_1$) is .1177832, with standard error .0049396.

A few questions for you:

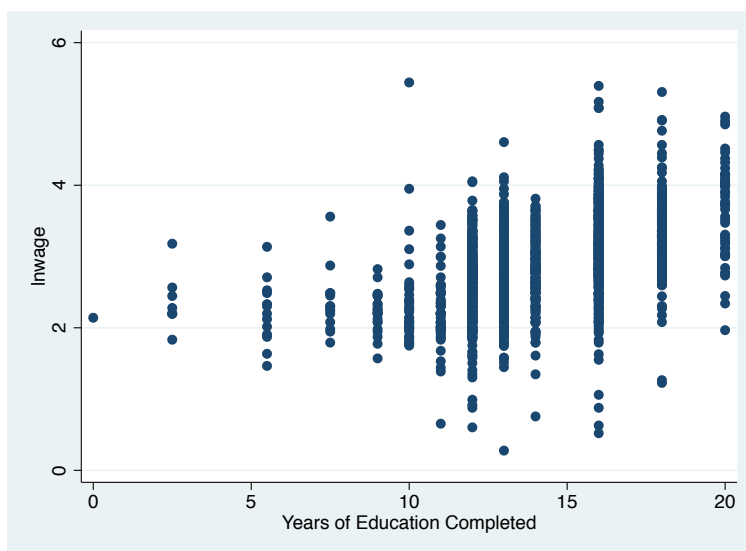
1. Is the slope statistically significant?
2. How do you interpret the slope?

3. What is the R^2 ? Briefly interpret this number.
4. What is the MSE?

Plotting relationships

When examining the relationship between two variables (such as education and wages), it is often a good idea to “visualize” the relationship by plotting the data. There are many benefits to doing so. For example, a plot will help us answer the questions: (a) Is the relationship between education and wages linear? (b) Is the regression estimate from above driven by outliers in the data?

The simplest way is `scatter lnwage educ`, which will plot each data point i , with observation i 's value of education on the x-axis and observation i 's value of `lnwage` on the y-axis. The graph looks like this:

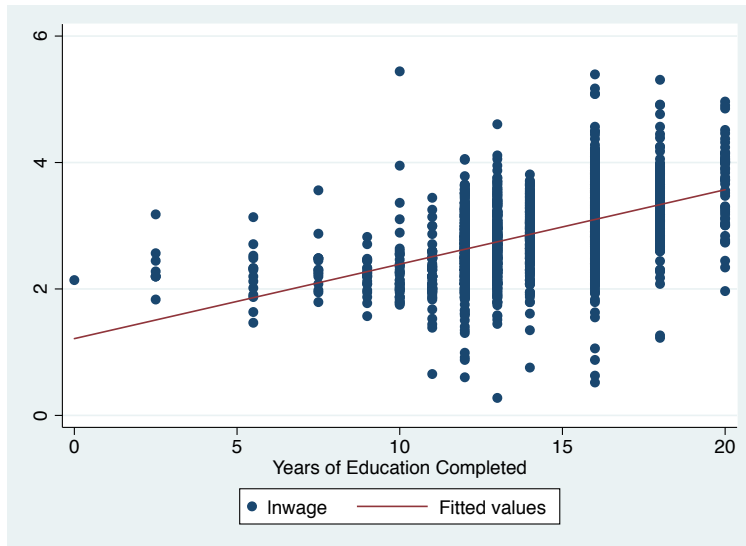


After creating a plot in STATA, you can save the plot as picture file on your computer using `graph export wagescatter1.png, replace`, which will store the graph in your working directory as `.png` image file that could then be inserted into, say, a word document.

We can also add the regression line from above to our scatter plot using the command:

```
graph twoway (scatter lnwage yrsed) (lfit lnwage yrsed)
```

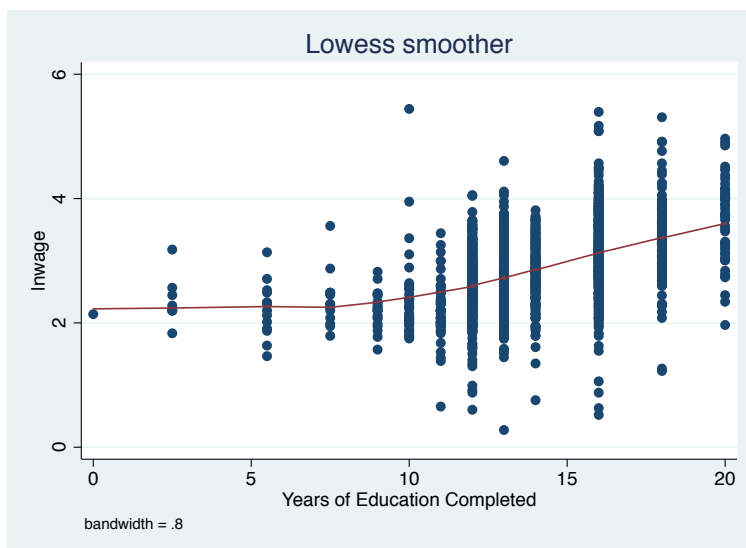
The two expressions in parentheses correspond to two different plots that will be shown on the same graph. The first part is our same scatter plot from above. The second part (`lfit`) will add the linear fitted line (same as our regression from above) to the graph:



A useful alternative to the above graph is to fit a general, nonlinear relationship between the two variables (this is called a *local linear regression*, you do not need to understand what this is right now). To get this more general fit, we can use:

```
lowess lnwage yrsed
```

This command generates the following graph:

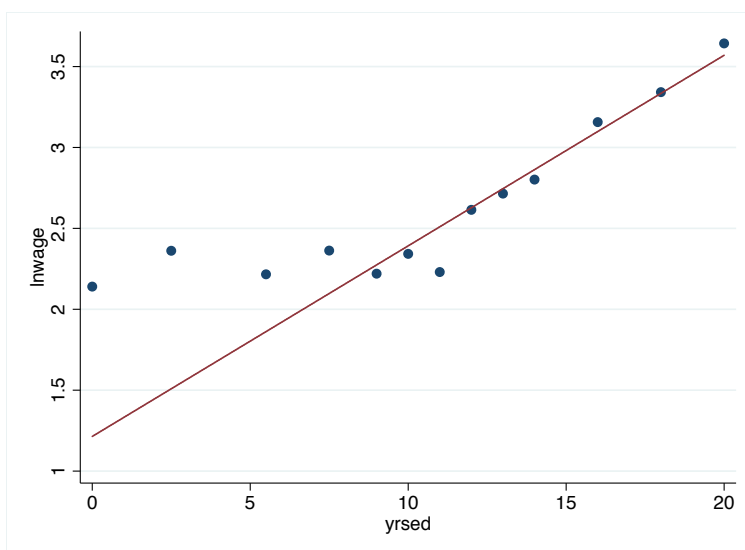


Once we have the nonlinear fit, does the relationship between years of education and $\ln(\text{wages})$ look linear?

Another very useful graphing command in STATA is called a binned scatter. The `binscatter` command chops the x variable up into bins (i.e., 5-6 years of education) and then plots the average value of y in that bin. In mathematical terms, we can think of `binscatter` as an empirical approximation of the conditional expectation function $E[y|x]$. The `binscatter` command is not natively installed in STATA, we first need to install it. To create a binned scatter plot, use the commands:

```
ssc install binscatter
binscatter lnwage yrsed
```

The first line installs the `binscatter` program. Once you have done this once, you never need to do it again. The command creates the following plot:



5 Stata programming

Entering calls into the command line one-by-one and examining output in the results panel is called using STATA *interactively*. This is a useful way to open STATA for the first time and get a sense of a few basic commands, but is a bad practice more generally. Going forward, we will instead use *programs* to analyze data. A STATA program, called a “.do file,” is really just a list of commands that will be passed to STATA line-by-line all at once. To create a new do-file, you can click the labeled **do-file editor** at the top of your STATA window.

There is a template do-file on canvas that you should download. Your do-file should always begin with the following three lines:

```
clear all
cap log close
set more off
```

The next step in every do-file is to initiate a “log” file. A log file prints all the output of your do-file into a text file that you can read after your program has finished running. A good practice is to name your log file the same thing as the your do-file, so that you can always keep track of which log file corresponds to which STATA program.

```
log using "LOGFILENAME.log", replace
```

After initiating your log file, you enter STATA commands line-by-line into your do-file. A simple do-file that only performs the `describe` and `summarize` commands would look like this:

```
clear all
cap log close
set more off
```

```
log using "LOGFILENAME.log", replace
```

```
describe
summarize
```

```
cap log close
```

The last line, `cap log close`, tells STATA to stop writing in the log-file. To execute a do-file, type `do DOFILENAME` into the command line.

Assignment for you: Download the template do-file and save it in the folder you created in the first step above. Rename the do-file as `stata_intro.do`. Modify the do-file so that it creates a log file called `stata_intro.log`. Then modify the do-file so that it executes the following commands:

1. Create the new variable `lnwage`.
2. Estimate the regression of `lnwage` on `yrsed`.
3. Construct a simple scatter plot fo `lnwage` against `yrsed`.
4. Store that plot as `wagescatter.png`.

Run the do-file and make sure it works. Examine the log-file and make sure it looks how you expect it to look. Make sure there is a new image file called `wagescatter.png` stored on your computer.

6 Additional exercises

Predicted values and residuals

Immediately after any regression command (`reg`), you can construct new variables containing either the predicted values, \hat{y} , or residuals, \hat{u} .

1. Add the command `predict yhat` to your program after `reg lnwage yrsed`. This constructs the predicted values. Take note that, in this case, `predict yhat` is equivalent in this case to using `gen yhat = 1.214163 + .1177*yrsed`. You can verify that these are equivalent directly in STATA if you want to get creative.
2. Construct the residuals by adding the command `predict uhat, resid`. Take note again that, in this specific case, this is equivalent to using the command `gen uhat = lnwage (1.214163 + -.1177* yrsed)`.
3. Now regress `uhat` on `yrsed`. Is there a statistically significant relationship between the residuals and years of education. Does this speak to whether the relationship estimated via `reg lnwage yrsed` is causal (in light of assumption SLR.4)? Why or why not?
4. Explain in words (no symbols) what needs to be true in order for the estimated slope to provide a causal estimate?
5. Now estimate the reverse regression, that is regress `yrsed` on `lnwage`. Is the slope just the reciprocal of the “forward” regression that you already ran?
6. How does the R^2 compare to the forward regression?