Heteroscedasticity

Econ 671

Purdue University

Is the OLS estimator still consistent? Let's see ...

$$\hat{\beta} = \beta + (X'X)^{-1}X'\epsilon$$
$$= \beta + \left(\frac{X'X}{n}\right)^{-1}\frac{X'\epsilon}{n}.$$

Therefore,

$$plim(\hat{\beta}) = \beta + plim\left[(X'X/n)^{-1}\right] plim\left(X'\epsilon/n\right).$$

From our previous lecture, we continue to assume that

$$plim[(X'X/n)^{-1}] = Q^{-1},$$

where Q^{-1} has finite elements.

Is the OLS estimator still consistent? Let us now consider the second term. Again, we can write:

$$n^{-1}X'\epsilon = \frac{1}{n}\sum_{i}x'_{i}\epsilon_{i}.$$

In our previous proofs, we employed a law of large numbers based on *iid* data together with mean independence to argue:

$$\frac{1}{n}\sum_{i}x'_{i}\epsilon_{i}\stackrel{p}{\rightarrow}E(x'_{i}\epsilon_{i})=0.$$

In this case, however, *the data are independent, but NOT identically distributed.*

We can, however, use a different LLN for independent observations. For example, Chebychev's LLN can be applied. This states:

Theorem

Let x_i be a series of independent, but not necessarily identically distributed random variables with $E(x_i) = \mu_i$ and $Var(x_i) = \sigma_i^2 < \infty$. If

$$\frac{1}{n^2}\sum_{i=1}^{\infty}\sigma_i^2\to 0$$

then

$$\overline{X}_n - \frac{1}{n} \sum_{i=1}^n \mu_i \stackrel{p}{\to} 0$$

In our regression application, provided the sequence of partial sums

doesn't grow too fast, we can use the theorem to show

 $n^{-1}X'\epsilon \to 0$

 $\sum_{i} \sigma_i^2 x_i' x_i$

so that the OLS estimator remains consistent.