

Solution to Exercise 6.2

***6.2** Consider a model similar to (3.20), but with error terms that are normally distributed:

$$y_t = \beta_1 + \beta_2 1/t + u_t, \quad u_t \sim \text{NID}(0, \sigma^2),$$

where $t = 1, 2, \dots, n$. If the true value of β_2 is β_2^0 and $\hat{\beta}_2$ is the OLS estimator, show that the limit in probability of $\hat{\beta}_2 - \beta_2^0$ is a normal random variable with mean 0 and variance $6\sigma^2/\pi^2$. In order to obtain this result, you will need to use the results that

$$\sum_{t=1}^{\infty} (1/t)^2 = \pi^2/6,$$

and that, if $s(n) = \sum_{t=1}^n (1/t)$, then $\lim n^{-1}s(n) = 0$ and $\lim n^{-1}s^2(n) = 0$.

Subtracting the sample mean \bar{y} from y_t and the sample mean $s(n)/n$ from $1/t$ is equivalent to regressing each of them on a constant term and taking the residuals. Thus, by the FWL Theorem, the OLS estimator $\hat{\beta}_2$ is the same as the estimator obtained by regressing $y_t - \bar{y}$ on $1/t - s(n)/n$, which is

$$\hat{\beta}_2 = \frac{\sum_{t=1}^n (y_t - \bar{y})(1/t - s(n)/n)}{\sum_{t=1}^n (1/t - s(n)/n)^2}. \quad (\text{S6.01})$$

Since the DGP is assumed to be (3.20) with $\beta_2 = \beta_2^0$,

$$y_t - \bar{y} = \beta_2^0 (1/t - s(n)/n) + u_t - \bar{u},$$

where \bar{u} denotes the mean of the u_t . Substituting for $y_t - \bar{y}$, (S6.01) can be rewritten as

$$\hat{\beta}_2 - \beta_2^0 = \frac{\sum_{t=1}^n (u_t - \bar{u})(1/t - s(n)/n)}{\sum_{t=1}^n (1/t - s(n)/n)^2}. \quad (\text{S6.02})$$

It is easy to see from (S6.02) that $\hat{\beta}_2 - \beta_2^0$ has mean zero. Since $1/t - s(n)/n$ is nonstochastic, the expectation of the numerator on the right-hand side is the same as the expectation of $u_t - \bar{u}$, which is zero. Because the denominator is also nonstochastic, the whole expression must have expectation zero. It must also be normally distributed, because the u_t are assumed to be normally distributed, and a weighted sum of normal random variables is itself normally distributed.

We must now determine the limit of the variance of the right-hand side of equation (S6.02) as $n \rightarrow \infty$. Since the mean is zero,

$$\text{Var}(\hat{\beta}_2 - \beta_2^0) = \text{E} \left(\frac{\sum_{t=1}^n (u_t - \bar{u})(1/t - s(n)/n)}{\sum_{t=1}^n (1/t - s(n)/n)^2} \right)^2. \quad (\text{S6.03})$$

The numerator of the expression on the right-hand side of this equation is the square of a sum of n terms. Since $E(u_t u_s) = 0$ for $t \neq s$, $\text{plim}(\bar{u}) = 0$. In addition, $\lim s(n)/n = 0$. Therefore, the expectation of this numerator is asymptotically equivalent to

$$E\left(\sum_{t=1}^n u_t^2 (1/t)^2\right) = \frac{1}{6} \pi^2 \sigma^2. \quad (\text{S6.04})$$

Because $\lim s(n)/n = 0$ and $\lim s^2(n)/n = 0$, the denominator of the expression on the right-hand side of equation (S6.03) is asymptotically equivalent to

$$\left(\lim_{n \rightarrow \infty} \sum_{t=1}^n (1/t)^2\right)^2 = \left(\frac{1}{6} \pi^2\right)^2. \quad (\text{S6.05})$$

Dividing the right-hand side of (S6.04) by the right-hand side of (S6.05) yields the result that

$$\text{Var}(\hat{\beta}_2 - \beta_2^0) \stackrel{a}{=} \frac{6\sigma^2}{\pi^2},$$

which is what we were required to prove.