Solution to Exercise 14.17

*14.17 Consider the simplest ADF testing regression

$$\Delta y_t = \beta' y_{t-1} + \delta \Delta y_{t-1} + e_t,$$

and suppose the data are generated by the standardized random walk (14.01), with $y_t = w_t$. If M_1 is the orthogonal projection matrix that yields residuals from a regression on the lagged dependent variable Δy_{t-1} , and if y_{-1} is the *n*-vector with typical element y_{t-1} , show that the expressions

$$\frac{1}{n}\sum_{t=1}^{n} (\boldsymbol{M}_{1}\boldsymbol{y}_{-1})_{t}\varepsilon_{t}$$
 and $\frac{1}{n}\sum_{t=1}^{n} y_{t-1}\varepsilon_{t}$

have the same probability limit as $n \to \infty$. Then derive the analogous result for the two expressions

$$\frac{1}{n^2} \sum_{t=1}^n (M_1 y_{-1})_t^2$$
 and $\frac{1}{n^2} \sum_{t=1}^n y_{t-1}^2$.

We can write

$$\boldsymbol{M}_{1}\boldsymbol{y}_{-1} = \boldsymbol{y}_{-1} - \Delta \boldsymbol{y}_{-1} (\Delta \boldsymbol{y}_{-1}^{\top} \Delta \boldsymbol{y}_{-1})^{-1} \Delta \boldsymbol{y}_{-1}^{\top} \boldsymbol{y}_{-1},$$

where Δy_{-1} is a vector with typical element Δy_{t-1} , which, under the specified DGP, is just ε_{t-1} . Therefore, $n^{-1} \sum_{t=1}^{n} (M_1 y_{-1})_t \varepsilon_t$ is equal to

$$\frac{1}{n}\sum_{t=1}^{n}y_{t-1}\varepsilon_t - \left(\frac{1}{n}\sum_{t=1}^{n}\varepsilon_{t-1}\varepsilon_t\right)\left(\frac{1}{n}\sum_{t=1}^{n}\varepsilon_{t-1}^2\right)^{-1}\left(\frac{1}{n}\sum_{t=1}^{n}\varepsilon_{t-1}y_{t-1}\right).$$
 (S14.17)

Because the ε_t are independent, with mean 0 and variance 1,

$$\lim_{n \to \infty} \frac{1}{n} \sum_{t=1}^{n} \varepsilon_{t-1} \varepsilon_t = 0 \quad \text{and} \quad \lim_{n \to \infty} \frac{1}{n} \sum_{t=1}^{n} \varepsilon_{t-1}^2 = 1.$$

Thus, provided the probability limit of the third factor in the second term of expression (S14.17) is finite, the whole second term must have a plim of zero. The plim of the third factor is

$$\lim_{n \to \infty} \frac{1}{n} \sum_{t=1}^{n} \varepsilon_{t-1} y_{t-1} = \lim_{n \to \infty} \frac{1}{n} \sum_{t=1}^{n-1} \varepsilon_t \sum_{s=1}^{t} \varepsilon_s = 1.$$

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Here we use the fact that ε_t is independent of ε_s for all s < t and equal to ε_s when s = t. Thus we conclude that the whole second term in expression (S14.17) has a plim of 0. Therefore,

$$\lim_{n \to \infty} \frac{1}{n} \sum_{t=1}^{n} (\boldsymbol{M}_1 \boldsymbol{y}_{-1})_t \varepsilon_t = \lim_{n \to \infty} \frac{1}{n} \sum_{t=1}^{n} y_{t-1} \varepsilon_t,$$

which is what we were required to prove.

For the second part of the question, we need the plim of $n^{-2} \sum_{t=1}^{n} (M_1 y_{-1})_t^2$, which can also be written as n^{-2} times the inner product of the vector $M_1 y_{-1}$ with itself. This inner product is

$$\boldsymbol{y}_{-1}^{\top} \boldsymbol{M}_{1} \boldsymbol{y}_{-1} = \sum_{t=1}^{n} y_{t-1}^{2} - \sum_{t=1}^{n} y_{t-1} \varepsilon_{t-1} \left(\sum_{t=1}^{n} \varepsilon_{t-1}^{2} \right)^{-1} \sum_{t=1}^{n} y_{t-1} \varepsilon_{t-1}.$$

We wish to show that the plim of n^{-2} times the second term on the right-hand side of this equation is equal to 0. The plim in question is

$$\frac{1}{n} \left(\lim_{n \to \infty} \frac{1}{n} \sum_{t=1}^{n} y_{t-1} \varepsilon_{t-1} \right) \left(\lim_{n \to \infty} \frac{1}{n} \sum_{t=1}^{n} \varepsilon_{t-1} \varepsilon_{t-1} \right)^{-1} \left(\lim_{n \to \infty} \frac{1}{n} \sum_{t=1}^{n} y_{t-1} \varepsilon_{t-1} \right).$$

We have already seen that each of the probability limits in parentheses is equal to 1. Therefore, the plim of the entire term is just $plim \frac{1}{n} = 0$. Thus we conclude that

$$\lim_{n \to \infty} \frac{1}{n^2} \sum_{t=1}^n (\boldsymbol{M}_1 \boldsymbol{y}_{-1})_t^2 = \lim_{n \to \infty} \frac{1}{n^2} \sum_{t=1}^n y_{t-1}^2,$$

which is what we were required to prove.