

Solution to Exercise 15.3

***15.3** Suppose the dependent variable \mathbf{y} is generated by the DGP

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta}_0 + \mathbf{u}, \quad \mathbf{u} \sim N(\mathbf{0}, \sigma_0^2 \mathbf{I}),$$

where the $n \times k$ matrix \mathbf{X} is independent of \mathbf{u} . Let \mathbf{z} be a vector that is not necessarily independent of \mathbf{u} , but is independent of $\mathbf{M}_\mathbf{X}\mathbf{u}$. Show that the t statistic on \mathbf{z} in the linear regression $\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + c\mathbf{z} + \mathbf{u}$ follows the Student's t distribution with $n - k - 1$ degrees of freedom.

The t statistic can be written as

$$\left(\frac{\mathbf{y}^\top \mathbf{M}_{\mathbf{X}, \mathbf{z}} \mathbf{y}}{n - k - 1} \right)^{-1/2} \frac{\mathbf{z}^\top \mathbf{M}_\mathbf{X} \mathbf{y}}{(\mathbf{z}^\top \mathbf{M}_\mathbf{X} \mathbf{z})^{1/2}}.$$

This is just the rightmost expression in equation (4.25) adapted to the test regression in which we are interested. In accordance with our usual notation, $\mathbf{M}_\mathbf{X}$ denotes the matrix that projects orthogonally onto $\mathcal{S}^\perp(\mathbf{X})$, and $\mathbf{M}_{\mathbf{X}, \mathbf{z}}$ denotes the matrix that projects orthogonally onto $\mathcal{S}^\perp(\mathbf{X}, \mathbf{z})$.

For the given DGP, $\mathbf{M}_\mathbf{X}\mathbf{y} = \mathbf{M}_\mathbf{X}\mathbf{X}\boldsymbol{\beta}_0 + \mathbf{M}_\mathbf{X}\mathbf{u} = \mathbf{M}_\mathbf{X}\mathbf{u}$, and the t statistic becomes

$$\left(\frac{\boldsymbol{\varepsilon}^\top \mathbf{M}_{\mathbf{X}, \mathbf{z}} \boldsymbol{\varepsilon}}{n - k - 1} \right)^{-1/2} \frac{\mathbf{z}^\top \mathbf{M}_\mathbf{X} \boldsymbol{\varepsilon}}{(\mathbf{z}^\top \mathbf{M}_\mathbf{X} \mathbf{z})^{1/2}}, \quad (\text{S15.04})$$

where $\boldsymbol{\varepsilon} \equiv \mathbf{u}/\sigma_0$. Notice that expression (S15.04) depends on $\boldsymbol{\varepsilon}$ only through the vector $\mathbf{M}_\mathbf{X}\boldsymbol{\varepsilon}$. This is the case because $\mathbf{M}_{\mathbf{X}, \mathbf{z}}\boldsymbol{\varepsilon} = \mathbf{M}_{\mathbf{X}, \mathbf{z}}\mathbf{M}_\mathbf{X}\boldsymbol{\varepsilon}$, which in turn is true by the result of Exercise 2.15.

We can now reason conditionally on \mathbf{z} . In order to compute the distribution of the statistic (S15.04) conditional on \mathbf{z} , we may proceed as though \mathbf{z} were deterministic, since \mathbf{z} is independent of $\mathbf{M}_\mathbf{X}\boldsymbol{\varepsilon}$ by hypothesis. But in Section 4.4, in the argument that follows equation (4.25), we showed that, when the regressors are exogenous, a statistic of the form (4.25) follows the $t(n - k)$ distribution, where k is the total number of regressors. Here, in the present notation, this means that, conditionally on \mathbf{z} , the statistic (S15.04) follows the $t(n - k - 1)$ distribution. This distribution does not depend on \mathbf{z} , and so it is also the unconditional distribution, as we wished to show.