## Solution to Exercise 15.21

\*15.21 Set up the OPG artificial regression for the Cox test of model  $H_1$  against  $H_2$  in (15.37), assuming IID normal errors. In particular, show that, in the notation of Exercise 15.20, the typical element of the test regressor (15.54) takes the form

$$\log\left(\frac{\hat{\sigma}_{1}^{2} + \hat{\sigma}_{a}^{2}}{\hat{\sigma}_{2}^{2}}\right) - \frac{\hat{u}_{2t}^{2}}{\hat{\sigma}_{2}^{2}} + \frac{\hat{\sigma}_{1}^{2} + (\boldsymbol{M}_{\boldsymbol{Z}}\boldsymbol{P}_{\boldsymbol{X}}\boldsymbol{y})_{t}^{2}}{\hat{\sigma}_{1}^{2} + \hat{\sigma}_{a}^{2}}, \qquad (15.91)$$

where  $\hat{u}_{2t}$  is a typical element of the vector  $M_Z y$ .

As we saw in the previous exercise, the terms that involve  $\ell_{1t}(\hat{\theta}_1)$  may be omitted when calculating Cox tests for linear regression models. Therefore, from (15.54), a typical element of the test regressor must be

$$\ell_{2t}(\hat{\boldsymbol{\theta}}_2) - \mathcal{E}_{\hat{\boldsymbol{\theta}}_1}(\ell_{2t}(\hat{\boldsymbol{\theta}}_2)).$$
(S15.44)

The  $t^{\text{th}}$  contribution to  $\ell_2(\hat{\theta}_2)$  is

$$\ell_{2t}(\hat{\theta}_2) = -\frac{1}{2}\log 2\pi - \frac{1}{2}\log \hat{\sigma}_2^2 - \frac{1}{2}\frac{\hat{u}_{2t}^2}{\hat{\sigma}_2^2}.$$
 (S15.45)

We saw in the solution to the previous exercise that, under a DGP in  $H_1$  with parameters  $\boldsymbol{\beta}$  and  $\sigma_1^2$ , the expectation of  $\log \hat{\sigma}_2^2$  is equal, with an error of order  $n^{-1}$  only, to  $\log(\sigma_1^2 + \sigma_a^2)$ , where  $\sigma_a^2$  is defined as plim  $n^{-1} \| \boldsymbol{M}_{\boldsymbol{Z}} \boldsymbol{X} \boldsymbol{\beta} \|^2$ . An entirely similar argument, based on a Taylor expansion, shows that the expectation of  $1/\hat{\sigma}_2^2$  is  $1/\sigma_a^2$  with error of order  $n^{-1}$  only.

The residual  $\hat{u}_{2t}$  is the  $t^{\text{th}}$  element of the vector  $M_Z y = M_Z u + M_Z X \beta$ . The expectation of  $\hat{u}_{2t}^2$  is therefore  $\sigma_1^2(1-h_t) + (M_Z X \beta)_t^2$ , where  $h_t$  is the  $t^{\text{th}}$  diagonal element of  $P_Z$ . A natural estimate of this expectation is

$$\hat{\sigma}_1^2 + (\boldsymbol{M_Z} \boldsymbol{P_X} \boldsymbol{y})_t^2$$

and thus a natural estimate of  $E(\ell_{2t}(\hat{\theta}_2))$  is

$$-\frac{1}{2}\log 2\pi - \frac{1}{2}\log(\hat{\sigma}_1^2 + \hat{\sigma}_a^2) - \frac{1}{2}\frac{\hat{\sigma}_1^2 + (M_Z P_X y)_t^2}{\hat{\sigma}_1^2 + \hat{\sigma}_a^2}.$$
 (S15.46)

Subtracting expression (S15.46) from expression (S15.45) and multiplying the result by 2 (which we do for simplicity because the scale of the test regressor has no effect on the value of the test statistic) yields

$$\begin{aligned} &-\log \hat{\sigma}_{2}^{2} - \frac{\hat{u}_{2t}^{2}}{\hat{\sigma}_{2}^{2}} + \log(\hat{\sigma}_{1}^{2} + \hat{\sigma}_{a}^{2}) + \frac{\hat{\sigma}_{1}^{2} + (\boldsymbol{M}_{\boldsymbol{Z}} \boldsymbol{P}_{\boldsymbol{X}} \boldsymbol{y})_{t}^{2}}{\hat{\sigma}_{1}^{2} + \hat{\sigma}_{a}^{2}} \\ &= \log\left(\frac{\hat{\sigma}_{1}^{2} + \hat{\sigma}_{a}^{2}}{\hat{\sigma}_{2}^{2}}\right) - \frac{\hat{u}_{2t}^{2}}{\hat{\sigma}_{2}^{2}} + \frac{\hat{\sigma}_{1}^{2} + (\boldsymbol{M}_{\boldsymbol{Z}} \boldsymbol{P}_{\boldsymbol{X}} \boldsymbol{y})_{t}^{2}}{\hat{\sigma}_{1}^{2} + \hat{\sigma}_{a}^{2}}, \end{aligned}$$

which is expression (15.91).

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