

since \mathbf{P}_1 plays the same role for the manifold \mathcal{R} as does \mathbf{P}_0 for \mathcal{X} . The LM statistic (3.48) is

$$\frac{1}{\tilde{\sigma}^2}(\mathbf{y} - \tilde{\mathbf{x}})^\top \tilde{\mathbf{P}}_X(\mathbf{y} - \tilde{\mathbf{x}}). \quad (5.76)$$

If we express the statistic in terms of quantities that are $O(1)$, we obtain

$$\frac{1}{\tilde{\sigma}^2} n^{-1/2}(\mathbf{y} - \tilde{\mathbf{x}})^\top \tilde{\mathbf{X}} (n^{-1} \tilde{\mathbf{X}}^\top \tilde{\mathbf{X}})^{-1} n^{-1/2} \tilde{\mathbf{X}}^\top (\mathbf{y} - \tilde{\mathbf{x}}). \quad (5.77)$$

Like $\hat{\mathbf{X}}_t$, $\tilde{\mathbf{X}}_t$ is asymptotically nonstochastic. Therefore, from (5.75),

$$\begin{aligned} n^{-1/2} \tilde{\mathbf{X}}^\top (\mathbf{y} - \tilde{\mathbf{x}}) &= n^{-1/2} \sum_{t=1}^n \tilde{\mathbf{X}}_t^\top \tilde{u}_t \\ &= n^{-1/2} \sum_{t=1}^n \mathbf{X}_{0t}^\top (\mathbf{M}_1 \mathbf{u})_t + o(1) \\ &= n^{-1/2} \sum_{t=1}^n (\mathbf{M}_1 \mathbf{X}_0)_t u_t + o(1) \\ &= n^{-1/2} \mathbf{X}_0^\top \mathbf{M}_1 \mathbf{u} + o(1). \end{aligned}$$

The matrix $n^{-1} \tilde{\mathbf{X}}^\top \tilde{\mathbf{X}}$ is asymptotically nonstochastic, just as $n^{-1} \hat{\mathbf{X}}^\top \hat{\mathbf{X}}$ is, and so the LM statistic (5.77) is asymptotically equivalent to

$$\mathbf{u}^\top \mathbf{M}_1 \mathbf{X}_0 (\sigma_0^2 \mathbf{X}_0^\top \mathbf{X}_0)^{-1} \mathbf{X}_0^\top \mathbf{M}_1 \mathbf{u} = \sigma_0^{-2} \mathbf{u}^\top \mathbf{M}_1 \mathbf{P}_0 \mathbf{M}_1 \mathbf{u}. \quad (5.78)$$

Since $\mathcal{S}(\mathbf{X}_1)$ is a subspace of $\mathcal{S}(\mathbf{X}_0)$, we have $\mathbf{P}_1 \mathbf{P}_0 = \mathbf{P}_0 \mathbf{P}_1 = \mathbf{P}_1$, from which it follows that $\mathbf{M}_1 \mathbf{P}_0 \mathbf{M}_1 = \mathbf{P}_0 - \mathbf{P}_1$. Expression (5.78) thus becomes

$$\sigma_0^{-2} \mathbf{u}^\top (\mathbf{P}_0 - \mathbf{P}_1) \mathbf{u} = \sigma_0^{-2} \mathbf{u}^\top \mathbf{P}_2 \mathbf{u}. \quad (5.79)$$

Comparison of (5.79) with (5.72) shows that the LM statistic is asymptotically equal to the Wald statistic. Thus it too is asymptotically $\chi^2(r)$ under the null hypothesis.

The third of the three test statistics discussed in Section 3.6 was the one based on the likelihood ratio principle, the pseudo- F statistic (3.50). Since we are interested in asymptotic results only, we rewrite it here in a form in which it should be asymptotically distributed as $\chi^2(r)$:

$$\frac{1}{s^2} (SSR(\tilde{\beta}) - SSR(\hat{\beta})) \quad (5.80)$$

and will (somewhat loosely) refer to it as the LR statistic. We have already seen that $s^2 \rightarrow \sigma_0^2$ as $n \rightarrow \infty$. It remains to show that $SSR(\tilde{\beta}) - SSR(\hat{\beta})$, when divided by σ_0^2 , is asymptotically $\chi^2(r)$. From (5.64), we have

$$\hat{\sigma}^2 = \frac{1}{n} \mathbf{u}^\top \mathbf{M}_0 \mathbf{u} + o(n^{-1}),$$