Proof of Wilk's phenomenon and Wald/Rao test

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In this note, we prove Theorem 7.3.1 in p.314,315 of the textbook.

(**Proof of (a)**&(b) of Theorem 7.3.1) For notational simplicity, let $\hat{\theta} \equiv \hat{\theta}_n^{\Omega}$. Since $\dot{\ell}(\hat{\theta}) = 0$, by Taylor's exapnsion,

$$\begin{split} 2(\ell(\hat{\theta}) - \ell(\theta_0)) &= 2\ell(\hat{\theta}) - 2[\ell(\hat{\theta}) + \dot{\ell}(\hat{\theta})(\theta_0 - \hat{\theta}) + \frac{1}{2}(\theta_0 - \hat{\theta})^t \ddot{\ell}(\theta^*)(\theta_0 - \hat{\theta})] \\ &= -(\theta_0 - \hat{\theta})^t \ddot{\ell}(\theta^*)(\theta_0 - \hat{\theta}) \\ &= \sqrt{n}(\theta_0 - \hat{\theta})^t \left[-\frac{1}{n} \ddot{\ell}(\theta^*) \right] \sqrt{n}(\theta_0 - \hat{\theta}), \end{split}$$

for some θ^* such that $|\theta^* - \theta_0| \leq |\hat{\theta} - \theta_0|$. Then, by Taylor's expansion

$$\frac{1}{n}\ddot{\ell}(\theta^*) = \frac{1}{n}\ddot{\ell}(\hat{\theta}) + \frac{1}{n}\ell^{(3)}(\theta^{**})(\theta^* - \theta_0),\tag{1}$$

for some θ^{**} such that $|\theta^{**} - \theta_0| \le |\theta^* - \theta_0|$. Let $R_n = \frac{1}{n}\ell^{(3)}(\theta^{**})(\theta^* - \theta_0)$. Fix $\epsilon > 0$ and choose k > 0. Then,

$$\mathbb{P}(|\theta_{0} - \theta^{*}| > \frac{\epsilon}{k}) \leq \mathbb{P}(|\hat{\theta} - \theta_{0}| > \frac{\epsilon}{k}) \stackrel{P}{\to} 0, \tag{2}$$

$$\mathbb{P}(|\frac{1}{n}\ell^{(3)}(\theta^{**})| > k) \leq \frac{1}{k}\mathbb{E}_{\theta_{0}}|\frac{1}{n}\ell^{(3)}(\theta^{**})|$$

$$\leq \frac{1}{k}\frac{1}{n}\sum_{i=1}^{n}\mathbb{E}_{\theta_{0}}M(X_{1})$$

$$= \frac{1}{k}\mathbb{E}_{\theta_{0}}M(X_{1}) \to 0 \text{ as } k \to \infty.$$
(3)

For the inequality in (2), we used (R6), which implies the consistency of $\hat{\theta}$. Also, for the second inequality in (3), we used (R7), and the limit in the last equation holds because $\mathbb{E}_{\theta_0}M(X_1) < \infty$ by (R7). Since

$$\mathbb{P}(|R_n| > \epsilon) \le \mathbb{P}(|\frac{1}{n}\ell^{(3)}(\theta^{**})| > k) + \mathbb{P}(|\theta_0 - \theta^*| > \epsilon/k),$$

(2) and (3) imply $R_n = o_p(1)$ by choosing sufficiently large k. Recall that $\frac{1}{n}\ddot{\ell}(\theta_0) \stackrel{P}{\to} -I(\theta_0)$ by weak law of large numbers (WLLN), and so $\frac{1}{n}\ddot{\ell}(\theta_0) = -I(\theta_0) + o_p(1)$. Thus, this together with $|R_n| = o_p(1)$ implies that

$$-\frac{1}{n}\ddot{\ell}(\theta^*) = I(\theta_0) + o_p(1) \tag{4}$$

by (1). Therefore,

$$2(\ell(\hat{\theta}) - \ell(\theta_0)) = \sqrt{n}(\theta_0 - \hat{\theta})^t \left[-\frac{1}{n}\ddot{\ell}(\theta^*) \right] \sqrt{n}(\theta_0 - \hat{\theta})$$

$$= \sqrt{n}(\theta_0 - \hat{\theta})^t \left[I(\theta_0) + o_p(1) \right] \sqrt{n}(\theta_0 - \hat{\theta})$$

$$= \sqrt{n}(\theta_0 - \hat{\theta})^t I(\theta_0) \sqrt{n}(\theta_0 - \hat{\theta}) + \sqrt{n}(\theta_0 - \hat{\theta})o_p(1) \sqrt{n}(\theta_0 - \hat{\theta}).$$

Recall that $\sqrt{n}(\hat{\theta} - \theta_0) \xrightarrow{d} N(0, I(\theta_0)^{-1})$ by (R0)-(R7), and so $\sqrt{n}(\hat{\theta} - \theta_0) = O_p(1)$. Hence,

$$\sqrt{n}(\theta_0 - \hat{\theta})o_p(1)\sqrt{n}(\theta_0 - \hat{\theta}) = o_p(1). \tag{5}$$

Also,

$$\sqrt{n}(\theta_0 - \hat{\theta})^t I(\theta_0) \sqrt{n}(\theta_0 - \hat{\theta}) = [I(\theta_0)^{\frac{1}{2}} \sqrt{n}(\theta_0 - \hat{\theta})]^t I(\theta_0)^{\frac{1}{2}} \sqrt{n}(\theta_0 - \hat{\theta})$$

and

$$I(\theta_0)^{\frac{1}{2}}\sqrt{n}(\theta_0-\hat{\theta})\stackrel{d}{\to} N(0,I_k),$$

by Δ -method. Since the function $g: \mathbb{R}^k \to \mathbb{R}$ defined by $g(\mathbf{u}) = \mathbf{u}^t \mathbf{u}$ is twice continuously differentiable, by continuous mapping theorem,

$$g(I(\theta_0)^{\frac{1}{2}}\sqrt{n}(\theta_0 - \hat{\theta})) = \sqrt{n}(\theta_0 - \hat{\theta})^t I(\theta_0)\sqrt{n}(\theta_0 - \hat{\theta}) \stackrel{d}{\to} \chi^2(k), \tag{6}$$

which proves Wald test ((b) of Theorem 7.3.1). Furthermore, (5) and (6) imply

$$2(\ell(\hat{\theta}) - \ell(\theta_0)) \stackrel{d}{\to} \chi^2(k)$$

by Slutsky's theorem, which proves Wilk's phenomenon ((a) of Theorem 7.3.1). Hence, if one approximates LRT by Wilk's phenomenon, the rejection rule is given by

Reject
$$H_0$$
 if $2(\ell(\hat{\theta}) - \ell(\theta_0)) > \chi_{\alpha}^2(k)$.

Similarly, if one conducts Wald test, the rejection rule is given by

Reject
$$H_0$$
 if $\sqrt{n}(\theta_0 - \hat{\theta})^t I(\theta_0) \sqrt{n}(\theta_0 - \hat{\theta}) > \chi_{\alpha}^2(k)$.

(Proof of (c) of Theorem 7.3.1) By Taylor's expansion,

$$\dot{\ell}(\theta_0) = \dot{\ell}(\hat{\theta}) + \ddot{\ell}(\hat{\theta})(\theta_0 - \hat{\theta}) + \frac{1}{2}(\hat{\theta} - \theta_0)^t \ell^{(3)}(\theta^*)(\hat{\theta} - \theta_0)
= \ddot{\ell}(\hat{\theta})(\theta_0 - \hat{\theta}) + \frac{1}{2}(\hat{\theta} - \theta_0)^t \ell^{(3)}(\theta^*)(\hat{\theta} - \theta_0)$$
(7)

for some θ^* such that $|\theta^* - \theta_0| \le |\hat{\theta} - \theta_0|$. The second equality holds because $\dot{\ell}(\hat{\theta}) = 0$ by (5). By (7),

$$\frac{1}{\sqrt{n}}\dot{\ell}(\theta_0) = \left(-\frac{1}{n}\ddot{\ell}(\hat{\theta})\right)\sqrt{n}(\hat{\theta} - \theta_0) + \frac{1}{2\sqrt{n}}(\hat{\theta} - \theta_0)^t\ell^{(3)}(\theta^*)(\hat{\theta} - \theta_0)$$

and so

$$\sqrt{n}(\hat{\theta} - \theta_0) = \left(-\frac{1}{n}\ddot{\ell}(\hat{\theta})\right)^{-1} \left(\frac{1}{\sqrt{n}}\dot{\ell}(\theta_0) - \frac{1}{2\sqrt{n}}(\hat{\theta} - \theta_0)^t \ell^{(3)}(\theta^*)(\hat{\theta} - \theta_0)\right). \tag{8}$$

Let $Q_n = \frac{1}{2\sqrt{n}}(\hat{\theta} - \theta_0)^t \ell^{(3)}(\theta^*)(\hat{\theta} - \theta_0)$. Fix $\epsilon > 0$ and take k > 0. Then,

$$\mathbb{P}(|Q_{n}| > \epsilon) \leq \mathbb{P}(|\frac{1}{2\sqrt{n}}(\hat{\theta} - \theta_{0})^{t}\ell^{(3)}(\theta^{*})(\hat{\theta} - \theta_{0})| > \epsilon)
\leq \mathbb{P}(|\sqrt{n}(\hat{\theta} - \theta_{0})| > k) + \mathbb{P}(|\frac{1}{2n}\ell^{(3)}(\theta^{*})(\hat{\theta} - \theta_{0})| > \epsilon/k)
= \mathbb{P}(|\sqrt{n}(\hat{\theta} - \theta_{0})| > k) + \mathbb{P}(|\frac{1}{n}\ell^{(3)}(\theta^{*})(\hat{\theta} - \theta_{0})| > 2\epsilon/k)
= \mathbb{P}(|\sqrt{n}(\hat{\theta} - \theta_{0})| > k) + \mathbb{P}(|\frac{1}{n}\ell^{(3)}(\theta^{*})| > k) + \mathbb{P}(|\hat{\theta} - \theta_{0}| > 2\epsilon/k^{2}).$$
(9)

As in the proof of $R_n = o_p(1)$ in the proof of (a) and (b) of Theorem 7.3.1, by taking sufficiently large k, $\mathbb{P}(|\frac{1}{n}\ell^{(3)}(\theta^*)| > k)$ can be made suffuciently small. Also, by (R5), $\mathbb{P}(|\hat{\theta} - \theta_0| > 2\epsilon/k^2)$ tends to 0

as $n \to \infty$, as $\hat{\theta}$ is a consistent estimator of θ_0 . Since $\sqrt{n}(\hat{\theta} - \theta_0)$ converges to $Z \sim N(0, I(\theta_0)^{-1})$ in distribution, $\sqrt{n}(\hat{\theta} - \theta_0) = O_p(1)$. Hence, $\mathbb{P}(|\sqrt{n}(\hat{\theta} - \theta_0)| > k)$ can be made sufficiently small if k is chosen to be sufficiently large. Therefore, by (9), $Q_n = o_p(1)$.

Also, as

$$-\frac{1}{n}\ddot{\ell}(\hat{\theta}) = -\frac{1}{n}\ddot{\ell}(\theta_0) - \frac{1}{n}\ell^{(3)}(\theta^{**})(\hat{\theta} - \theta_0),$$
$$\frac{1}{n}\ell^{(3)}(\theta^{**}) = o_p(1),$$

using the similar argument in the proof of $R_n = o_p(1)$ in the proof of (a) and (b) of Theorem 7.3.1. Since $\hat{\theta}$ is a consistent estimator of θ_0 by (R5), $\hat{\theta} - \theta_0 = o_p(1)$. Also, as $-\frac{1}{n}\ddot{\ell}(\theta_0)$ converges in probability to $I(\theta_0)$ by WLLN, one can see that

$$-\frac{1}{n}\ddot{\ell}(\hat{\theta}) = I(\theta_0) + o_p(1). \tag{10}$$

By (8), (9) and (10),

$$\begin{split} \sqrt{n}(\hat{\theta} - \theta_0) &= (I(\theta_0) + o_p(1))^{-1} (\frac{1}{\sqrt{n}} \dot{\ell}(\theta_0) + o_p(1)) \\ &= (I(\theta_0)^{-1} + o_p(1)) (\frac{1}{\sqrt{n}} \dot{\ell}(\theta_0) + o_p(1)) \\ &= I(\theta_0)^{-1} \frac{1}{\sqrt{n}} \dot{\ell}(\theta_0) + o_p(1). \end{split}$$

Recall that under (R0)-(R7),

$$\sqrt{n}(\hat{\theta} - \theta_0)^t I(\theta_0) \sqrt{n}(\hat{\theta} - \theta_0) \xrightarrow{d} N(0, I(\theta_0)^{-1}).$$

Substituting $\sqrt{n}(\hat{\theta} - \theta_0) = I(\theta_0)^{-1} \frac{1}{\sqrt{n}} \dot{\ell}(\theta_0) + o_p(1)$ gives

$$\sqrt{n}(\hat{\theta} - \theta_0)^t I(\theta_0) \sqrt{n}(\hat{\theta} - \theta_0) = \left(I(\theta_0)^{-1} \frac{1}{\sqrt{n}} \dot{\ell}(\theta_0) + o_p(1) \right)^t I(\theta_0) \left(I(\theta_0)^{-1} \frac{1}{\sqrt{n}} \dot{\ell}(\theta_0) + o_p(1) \right) \\
= \frac{1}{n} \dot{\ell}(\theta_0)^t I(\theta_0)^{-1} \dot{\ell}(\theta_0) + o_p(1).$$

Again, by Slutsky's theorem,

$$\frac{1}{n}\dot{\ell}(\theta_0)^t I(\theta_0)^{-1}\dot{\ell}(\theta_0) \stackrel{d}{\to} N(0, I(\theta_0)^{-1}).$$

So if one conducts Rao test, the rejection rule is given by

Reject
$$H_0$$
 if $\frac{1}{n}\dot{\ell}(\theta_0)^t I(\theta_0)^{-1}\dot{\ell}(\theta_0) > \chi_{\alpha}^2(k)$.