

# Estimation with Overidentifying Inequality Moment Conditions – Technical Appendix

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August 27, 2008

(Not intended to appear in print)

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# 1 Consistency

For a precise definition of the notation see the main text.

## 1.1 Assumptions

- Assumption 1** (a)  $X_i, i = 1, \dots, n$  are strictly stationary on a probability space  $(\Omega_n, \mathcal{F}_n, P_n)$ ;  
 (b)  $\Theta$  is an  $m$ -dimensional compact subset of  $\mathbb{R}^m$ , where  $m \leq h_1$ ,  $\theta_{n,0} \rightarrow \theta_0$ ,  $\theta_{n,0} \in \Theta \forall n$ , and  $\theta_0 \in \Theta$ ;  
 (c)  $g(x, \theta)$  is continuous at each  $\theta \in \Theta$  with probability one;  
 (d)  $\mathbb{E}_n[g_1(X_i, \theta_{n,0})] = 0$ , and  $\inf_n \|\mathbb{E}_n[g_1(X_i, \theta)]\| > 0$  for  $\theta \neq \theta_{n,0}$ ;  
 (e)  $\nu_{n,0} \rightarrow \nu_0$  and  $\sqrt{n}\nu_{n,0} \rightarrow u_0 \in [0, \infty]^{h_2}$ ;  
 (f)  $\mathbb{E}_n[g(X_i, \theta_{n,0})g(X_i, \theta_{n,0})'] \rightarrow J$  is non-singular;  
 (g)  $Z_n = O_p(1)$ ;  
 (h)  $\mathbb{V} = \{\nu \in \mathbb{R}^{h_2} : \nu \geq 0 \text{ and } \|\nu\| \leq K\}$ ,  $\nu_{n,0} \in \mathbb{V} \forall n$ , and  $\nu_0$  lies in the interior of  $\mathbb{V}$ ;  
 (i)  $\mathbb{E}_n \left[ \sup_{\theta \in \Theta} \|g(X_i, \theta)\|^\alpha \right] \leq K < \infty$  for some  $\alpha > 2$ ;  
 (j) for any  $\theta$  and  $\theta^*$ ,  $\|g(X_i, \theta) - g(X_i, \theta^*)\| \leq L(X_i)l(\|\theta - \theta^*\|)$ , for some measurable function  $L$  of  $X_i$  such that  $\sup_n \mathbb{E}_n(L(X_i)) < \infty$ , and  $l(y) \downarrow 0$  as  $y \downarrow 0$ .

## 1.2 Main Results

**Theorem 1** Suppose that Assumption 1 is satisfied. Then  $\hat{\theta}_n - \theta_{n,0} \xrightarrow{P} 0$  and  $\hat{\nu}_n - \nu_{n,0} \xrightarrow{P} 0$ . Moreover,  $\hat{\lambda}(\hat{\theta}_n, \hat{\nu}_n) \xrightarrow{P} 0$ .

Throughout this appendix we are frequently using the following results. Notice that Assumption 1 implies that

$$\max_{1 \leq i \leq n} \sup_{\theta \in \Theta} \|g(X_i, \theta)\| = O_p(n^{1/\alpha}), \quad (1)$$

$$\frac{1}{n} \sum_{i=1}^n \sup_{\theta \in \Theta} \|g(X_i, \theta)\|^\alpha = O_p(1), \quad (2)$$

$$\sup_{\theta \in \Theta} \left\| \frac{1}{n} \sum_{i=1}^n \{g(X_i, \theta)g(X_i, \theta)' - \mathbb{E}[g(X_i, \theta)g(X_i, \theta)']\} \right\| = o_p(1). \quad (3)$$

According to Assumptions 1 and 1,

$$\mathbb{E}[g^{(1)}(X_i, \theta)] \text{ and } \mathbb{E}[g_j^{(2)}(X_i, \theta)] \quad (4)$$

are equicontinuous uniformly in  $\theta$ , and

$$\begin{aligned} \sup_{\theta \in \Theta} \left\| \frac{1}{n} \sum_{i=1}^n \{g^{(1)}(X_i, \theta) - \mathbb{E}[g^{(1)}(X_i, \theta)]\} \right\| &= o_p(1) \\ \sup_{\theta \in \Theta} \left\| \frac{1}{n} \sum_{i=1}^n \{g_j^{(2)}(X_i, \theta) - \mathbb{E}[g_j^{(2)}(X_i, \theta)]\} \right\| &= o_p(1) \text{ for all } j = 1, \dots, h. \end{aligned} \quad (5)$$

(See, for instance, Andrews (1992)).

**Proof of Theorem 1:** We have to show that for any  $\delta > 0$

$$\lim_{n \rightarrow \infty} P \left\{ \hat{\theta}_n \in \mathbb{B}(\theta_{n,0}, \delta), \hat{\nu}_n \in \mathbb{B}(\nu_{n,0}, \delta) \right\} = 1,$$

where

$$\mathbb{B}(\theta, \delta) = \{ \tilde{\theta} \in \Theta \mid \|\theta - \tilde{\theta}\| < \delta \}, \quad \mathbb{B}(\nu, \delta) = \{ \tilde{\nu} \in \mathbb{V} \mid \|\nu - \tilde{\nu}\| < \delta \}.$$

Define

$$\Theta_0^c = \Theta \cap \mathbb{B}(\theta_{n,0}, \delta)^c \quad \text{and} \quad N_0^c = \mathbb{V} \cap \mathbb{B}(\nu_{n,0}, \delta)^c.$$

To simplify the notation we omit the subscript  $n$  from the sets  $\Theta_0^c$  and  $N_0^c$ . Recall that according to Assumption 1(i), the constant  $\alpha > 2$  is such that  $\mathbb{E}[\sup_{\theta \in \Theta} \|g(X_i, \theta)\|^\alpha] < K$ . We show the following two statements are true: (i) For a given  $\varepsilon, \delta > 0$  and  $\zeta$  such that  $\frac{1}{\alpha} < \zeta < \frac{1}{2}$ , there exist positive constants  $\eta$  and  $\kappa$  and  $\bar{n}$  such that for  $n \geq \bar{n}$

$$P \left\{ \bar{G}_n^*(\theta_{n,0}, \nu_{n,0}) \geq n^{-\zeta - \kappa} \eta \right\} < \frac{\varepsilon}{2}, \quad (6)$$

where

$$\bar{G}_n^*(\theta_{n,0}, \nu_{n,0}) = \max_{\lambda \in \hat{\Lambda}_n(\theta_{n,0})} G_n^*(\theta_{n,0}, \nu_{n,0}, \lambda),$$

and (ii)

$$P \left\{ \min_{\theta \in \Theta_0^c, \nu \in N_0^c} \bar{G}_n^*(\theta, \nu) \leq n^{-\zeta} \eta \right\} < \frac{\varepsilon}{2}. \quad (7)$$

Then, from (6) and (7) we deduce that there exists an  $\eta > 0$  such that for  $n \geq \bar{n}$ :

$$\begin{aligned} & P \left\{ \hat{\theta}_n \in \mathbb{B}(\theta_{n,0}, \delta), \hat{\nu}_n \in \mathbb{B}(\nu_{n,0}, \delta) \right\} \\ & \geq P \left\{ \bar{G}_n^*(\theta_{n,0}, \nu_{n,0}) < n^{-\zeta - \kappa} \eta, \min_{\theta \in \Theta_0^c, \nu \in N_0^c} \bar{G}_n^*(\theta, \nu) > n^{-\zeta} \eta \right\} \geq 1 - \varepsilon. \end{aligned}$$

**Proof of (i):** By Lemma 2  $\bar{G}_n^*(\theta_{n,0}, \nu_{n,0}) \leq O_p(1/n)$ . Choose  $\kappa > 0$  such that  $\zeta + \kappa < 1$ . Then

$$n^{\zeta + \kappa} \bar{G}_n^*(\theta_{n,0}, \nu_{n,0}) \leq O_p(n^{\zeta + \kappa - 1}) = o_p(1)$$

as required.

**Proof of (ii):** To obtain a lower bound for  $\bar{G}_n^*(\theta, \nu)$  we will evaluate the function  $G_n^*(\theta, \nu, \lambda)$  at  $\lambda = n^{-\zeta} u(\theta, \nu)$ , where the function  $u(\theta, \nu)$  is defined as

$$u(\theta, \nu) = \begin{cases} 0 & \text{if } \theta = \theta_{n,0}, \nu = \nu_{n,0} \\ \frac{\mathbb{E}[g(X_i, \theta)] - M' \nu}{\|\mathbb{E}[g(X_i, \theta)] - M' \nu\|} & \text{otherwise} \end{cases}$$

such that  $\|u(\theta, \nu)\| \leq 1$ .

Moreover, we truncate the function  $g(x, \theta)$  as follows. Since  $\alpha > 2$ , we can choose a positive constant  $\xi$  such that

$$\frac{1}{\alpha^2} < \xi < \frac{1}{2\alpha}.$$

Let

$$\mathcal{X}_n = \left\{ x : \sup_{\theta \in \Theta} \|g(x, \theta)\| \leq n^\xi \right\} \quad \text{and} \quad g_n(x, \theta) = I \{x \in \mathcal{X}_n\} g(x, \theta).$$

We then replace the terms

$$\ln(1 + \lambda' g(x, \theta)) - \lambda' M' \nu$$

in the definition of the objective function  $G_n^*(\theta, \nu, \lambda)$  by

$$q_n(x, \theta, \nu) = \ln(1 + n^{-\zeta} u'(\theta, \nu) g_n(x, \theta)) - n^{-\zeta} u'(\theta, \nu) M' \nu.$$

In what follows, we deduce the required result for (ii) by showing that

$$\text{(ii)-(a): } \min_{\theta \in \Theta_0^c, \nu \in N_0^c} \frac{1}{n} \sum_{i=1}^n q_n(X_i, \theta, \nu) \leq \min_{\theta \in \Theta_0^c, \nu \in N_0^c} \bar{G}_n^*(\theta, \nu) + o_p(n^{-\zeta})$$

and

$$\text{(ii)-(b): } P \left\{ \min_{\theta \in \Theta_0^c, \nu \in N_0^c} \frac{1}{n} \sum_{i=1}^n q_n(X_i, \theta, \nu) < n^{-\zeta} \eta \right\} \leq \frac{\varepsilon}{2}.$$

**Proof of (ii)-(a):** Notice that  $n^{-\zeta} u(\theta, \nu) \in \Lambda_n^\zeta \subset \cap_{\theta \in \Theta} \hat{\Lambda}_n(\theta)$  w.p.a.1 by Lemma 1. Then, by Lemma 4 and by the definition of  $\hat{\lambda}_n(\theta, \nu)$ ,

$$\begin{aligned} & \min_{\theta \in \Theta_0^c, \nu \in N_0^c} \frac{1}{n} \sum_{i=1}^n q_n(X_i, \theta, \nu) \\ &= \min_{\theta \in \Theta_0^c, \nu \in N_0^c} \left[ \frac{1}{n} \sum_{i=1}^n \ln(1 + n^{-\zeta} u'(\theta, \nu) g(X_i, \theta)) - n^{-\zeta} u'(\theta, \nu) M' \nu \right] + o_p(n^{-\zeta}) \\ &\leq \min_{\theta \in \Theta_0^c, \nu \in N_0^c} \left[ \frac{1}{n} \sum_{i=1}^n \ln(1 + \hat{\lambda}'_n(\theta, \nu) g(X_i, \theta)) - \hat{\lambda}'_n(\theta, \nu) M' \nu \right] + o_p(n^{-\zeta}) \\ &= \min_{\theta \in \Theta_0^c, \nu \in N_0^c} \bar{G}_n^*(\theta, \nu) + o_p(n^{-\zeta}), \end{aligned}$$

as required.

**Proof of (ii)-(b):** A second-order Taylor expansion of  $q_n$  around  $u(\theta, \nu) = 0$  yields

$$n^\zeta q_n(x, \theta, \nu) = u(\theta, \nu)' (g_n(x, \theta) - M' \nu) - \frac{1}{2} \frac{n^{-\zeta} u'(\theta, \nu) g_n(x, \theta) g_n(x, \theta)' u(\theta, \nu)}{(1 + n^{-\zeta} u'_*(\theta, \nu) g_n(x, \theta))^2}, \quad (8)$$

where  $u_*(\theta, \nu)$  lies between zero and  $u(\theta, \nu)$ . The second-order term of the Taylor approximation (8) can be bounded as follows. For given  $x, \theta$ , and  $\nu$

$$\sup_{\theta \in \Theta, \nu} \left| n^{-\zeta} u'_*(\theta, \nu) g_n(x, \theta) \right| \leq n^{-\zeta} \sup_{\theta \in \Theta} \|g_n(x, \theta)\| \leq n^{-\zeta + \xi} \leq n^{-\zeta/2}$$

since  $\xi < \frac{1}{2\alpha} < \frac{\zeta}{2}$ . Therefore,

$$\sup_{\theta \in \Theta, \nu} n^{-\zeta} \frac{u(\theta, \nu)' g_n(x, \theta) g_n(x, \theta)' u(\theta, \nu)}{(1 + n^{-\zeta} u'_*(\theta, \nu) g_n(x, \theta))^2} \leq \sup_{\theta \in \Theta, \nu} n^{-\zeta} \frac{\|g_n(x, \theta)\|^2 \|u(\theta, \nu)\|^2}{(1 - n^{-\zeta/2})^2} \leq n^{-\zeta + 2\xi} = o(1). \quad (9)$$

Now consider the expected value of  $n^\zeta q_n(x, \theta, \nu)$ . From (8), (9), and by the dominated convergence theorem, we have

$$\begin{aligned} n^\zeta \mathbb{E}[q_n(X_i, \theta, \nu)] &= u'(\theta, \nu)(\mathbb{E}[g_n(X_i, \theta)] - M'\nu) + o(1) \\ &= \begin{cases} o(1) & \text{if } \theta = \theta_0, \nu = \nu_{n,0} \\ \|\mathbb{E}[g(X_i, \theta)] - M'\nu\| + o(1) > 0 & \text{otherwise} \end{cases}. \end{aligned} \quad (10)$$

The  $o(1)$  terms absorb the second-order term of the Taylor approximation and the discrepancy between  $\mathbb{E}[g_n(X, \theta)]$  and  $\mathbb{E}[g(X, \theta)]$ , which vanishes as  $\mathcal{X}_n$  expands. From (10) and the monotone convergence theorem we can deduce that

$$\lim_{n \rightarrow \infty} n^\zeta \lim_{\delta \downarrow 0} \mathbb{E} \left[ \inf_{\theta^* \in \mathbb{B}(\theta, \delta), \nu^* \in \mathbb{B}(\nu, \delta)} q_n(X_i, \theta^*, \nu^*) \right] \begin{cases} = 0 & \text{if } \theta = \theta_{n,0}, \nu = \nu_{n,0} \\ > 0 & \text{otherwise} \end{cases}.$$

Since  $\Theta$  and  $\mathbb{V}$  are compact by assumption, the sets  $\Theta \cap \mathbb{B}(\theta_{n,0}, \delta)^c$  and  $\mathbb{V} \cap \mathbb{B}(\nu_{n,0}, \delta)^c$  are compact. We can cover both  $\Theta \cap \mathbb{B}(\theta_{n,0}, \delta)^c$  and  $\mathbb{V} \cap \mathbb{B}(\nu_{n,0}, \delta)^c$  with  $\Theta_j = \mathbb{B}(\theta_j, \delta_j)$  and  $N_j = \mathbb{B}(\nu_j, \delta_j)$ 's,  $j = 1, \dots, J$  taking each  $\delta_j$  small enough such there exist  $\eta_j$ 's such that

$$n^\zeta \mathbb{E} \left[ \inf_{\theta \in \Theta_j, \nu \in N_j} q_n(X_i, \theta, \nu) \right] \geq 2\eta_j, \quad n \geq n_j \quad (11)$$

for some positive numbers  $\eta_j = \eta_j(\delta)$ ,  $j = 1, \dots, J$ . By the WLLN<sup>1</sup> and (11), for a given  $\varepsilon > 0$ , we can find  $\bar{n}_j$ 's such that  $n \geq \bar{n}_j$  implies that

$$\begin{aligned} \frac{\varepsilon}{2J} &\geq P \left\{ \left| \frac{1}{n} \sum_{i=1}^n n^\zeta \inf_{\theta \in \Theta_j, \nu \in N_j} q_n(X_i, \theta, \nu) - \mathbb{E} \left[ n^\zeta \inf_{\theta \in \Theta_j, \nu \in N_j} q_n(X_i, \theta, \nu) \right] \right| > \eta_j \right\} \\ &\geq P \left\{ \frac{1}{n} \sum_{i=1}^n \inf_{\theta \in \Theta_j, \nu \in N_j} q_n(X_i, \theta, \nu) < \mathbb{E} \left[ \inf_{\theta \in \Theta_j, \nu \in N_j} q_n(X_i, \theta, \nu) \right] - n^{-\zeta} \eta_j \right\} \\ &\geq P \left\{ \frac{1}{n} \sum_{i=1}^n \inf_{\theta \in \Theta_j, \nu \in N_j} q_n(X_i, \theta, \nu) < n^{-\zeta} \eta_j \right\} \\ &\geq P \left\{ \inf_{\theta \in \Theta_j, \nu \in N_j} \frac{1}{n} \sum_{i=1}^n q_n(X_i, \theta, \nu) < n^{-\zeta} \eta_j \right\} \end{aligned}$$

for  $j = 1, \dots, J$ . Now let letting  $\eta = \min \{\eta_1, \dots, \eta_J\}$  and  $\bar{n} = \max_{j=1, \dots, J} \bar{n}_j$ , we have for  $n \geq \bar{n}$

$$\begin{aligned} &P \left\{ \min_{\theta \in \Theta_0^c, \nu \in N_0^c} \frac{1}{n} \sum_{i=1}^n q_n(X_i, \theta, \nu) < n^{-\zeta} \eta \right\} \\ &\leq P \left\{ \min_{j=1, \dots, J} \left\{ \inf_{\theta \in \Theta_j, \nu \in N_j} \frac{1}{n} \sum_{i=1}^n q_n(X_i, \theta, \nu) \right\} < n^{-\zeta} \eta \right\} \\ &\leq \sum_{j=1}^J P \left\{ \inf_{\theta \in \Theta_j, \nu \in N_j} \frac{1}{n} \sum_{i=1}^n q_n(X_i, \theta, \nu) < n^{-\zeta} \eta_j \right\} \leq \frac{\varepsilon}{2}, \end{aligned}$$

<sup>1</sup>Notice that

$$\mathbb{E} \left[ \left( n^\zeta \inf_{\theta \in \Theta_j, \nu \in N_j} q_n(X_i, \theta, \nu) \right)^2 \right] \leq \mathbb{E} \left[ \sup_{\theta \in \Theta} 2 \|g(X_i, \theta)\|^2 \right] + 2K + n^{-2\zeta + 4\varepsilon} < \infty. \quad (12)$$

as required part (ii)-(b).

Combining (ii)-(a) and (ii)-(b) we have

$$P \left\{ \min_{\theta \in \Theta_0^c, \nu \in N_0^c} \bar{G}_n^*(\theta, \nu) < n^{-\zeta} \eta \right\} \leq \frac{\varepsilon}{2},$$

as required for (ii).

Since  $\hat{\theta}_n - \theta_{n,0} \xrightarrow{p} 0$  and  $\hat{\nu}_n - \nu_{n,0} \xrightarrow{p} 0$  we can deduce from Lemmas 2 and 3 that  $\hat{\lambda}(\hat{\theta}_n, \hat{\nu}_n) \xrightarrow{p} 0$ . ■

### 1.3 Technical Lemmas

**Lemma 1** *Suppose that Assumption 1 is satisfied. Then,*

- (i)  $\sup_{\theta \in \Theta, \lambda \in \Lambda_n^\zeta, 1 \leq i \leq n} |\lambda' g(X_i, \theta)| \xrightarrow{p} 0$ ,
- (ii)  $\Lambda_n^\zeta \subseteq \bigcap_{\theta \in \Theta} \hat{\Lambda}_n(\theta)$  w.p.a. 1.

**Proof of Lemma 1:** See proof of Lemma A1 in Newey and Smith (2004). ■

**Lemma 2** *Suppose that Assumption 1 is satisfied. Let  $\bar{\theta} \in \Theta$  and  $\bar{\nu} \geq 0$  be sequences such that  $\bar{\theta} - \theta_{n,0} \xrightarrow{p} 0$ , and  $\bar{\nu} - \nu_{n,0} \xrightarrow{p} 0$ . Moreover,  $\frac{1}{\sqrt{n}} \sum_{i=1}^n g_1(X_i, \bar{\theta}) = O_p(1)$  and  $\frac{1}{\sqrt{n}} \sum_{i=1}^n (g_2(X_i, \bar{\theta}) - \bar{\nu}) = O_p(1)$ . Then,*

- (i)  $\hat{\lambda}(\bar{\theta}, \bar{\nu})$  exists w.p.a. 1,
- (ii)  $\hat{\lambda}(\bar{\theta}, \bar{\nu}) = O_p(n^{-1/2})$ ,
- (iii)  $G_n^*(\bar{\theta}, \bar{\nu}, \hat{\lambda}(\bar{\theta}, \bar{\nu})) \leq O_p\left(\frac{1}{n}\right)$ .

**Proof of Lemma 2:** The proof is similar to that of Lemma A2 in Newey and Smith (2004).

**Proof of (i):** Define

$$\tilde{\lambda}(\bar{\theta}, \bar{\nu}) = \arg \max_{\lambda \in \Lambda_n^\zeta} G_n^*(\bar{\theta}, \bar{\nu}, \lambda)$$

Since  $\Lambda_n^\zeta$  is compact and  $\ln(1 + \lambda' g(X_i, \bar{\theta})) - \bar{\nu}' M \lambda$  is continuous and strictly concave in  $\lambda$  the optimal solution  $\tilde{\lambda}(\bar{\theta}, \bar{\nu})$  exists and is unique. Statement (i) then follows from Lemma 1.

**Proof of (ii) and (iii):** Write  $\bar{g}_i = g(X_i, \bar{\theta})$ . For some constant  $C$

$$\begin{aligned}
0 = G_n^*(\bar{\theta}, \bar{\nu}, 0) &\leq G_n^*(\bar{\theta}, \bar{\nu}, \tilde{\lambda}(\bar{\theta}, \bar{\nu})) \\
&= \frac{1}{n} \sum_{i=1}^n \ln \left( 1 + \tilde{\lambda}'(\bar{\theta}, \bar{\nu}) \bar{g}_i \right) - \bar{\nu}' M \tilde{\lambda}(\bar{\theta}, \bar{\nu}) \\
&= \tilde{\lambda}'(\bar{\theta}, \bar{\nu}) \left( \frac{1}{n} \sum_{i=1}^n \bar{g}_i - M' \bar{\nu} \right) - \frac{1}{2} \tilde{\lambda}''(\bar{\theta}, \bar{\nu}) \left( \frac{1}{n} \sum_{i=1}^n \frac{\bar{g}_i \bar{g}_i'}{(1 + \lambda'_* \bar{g}_i)^2} \right) \tilde{\lambda}(\bar{\theta}, \bar{\nu}) \\
&\leq \tilde{\lambda}'(\bar{\theta}, \bar{\nu}) \left( \frac{1}{n} \sum_{i=1}^n \bar{g}_i - M' \bar{\nu} \right) - \frac{C}{4} \tilde{\lambda}'(\bar{\theta}, \bar{\nu}) \tilde{\lambda}(\bar{\theta}, \bar{\nu}),
\end{aligned}$$

where  $\lambda_*$  lies on the line joining  $\tilde{\lambda}(\bar{\theta}, \bar{\nu})$  and 0. The last inequality holds because

$$\max_{1 \leq i \leq n} |\lambda'_* \bar{g}_i| = o_p(1)$$

according to Lemma 1 and  $\frac{1}{n} \sum_{i=1}^n \bar{g}_i \bar{g}_i'$  converges in probability to  $J$ , a positive definite matrix, by (3) and Assumption 1(f). The remainder of the proof follows the proof of Lemma A2 in Newey and Smith (2004). ■

**Lemma 3** *Suppose Assumption 1 is satisfied. Then,*

$$\frac{1}{\sqrt{n}} \sum_{i=1}^n \left[ g(X_i, \hat{\theta}) - M' \hat{\nu} \right] = O_p(1).$$

**Proof of Lemma 3:** The proof is similar to that of Lemma A.3 in Newey and Smith (2004).

Let  $\hat{g}_i = g(X_i, \hat{\theta}) - M' \hat{\nu}$  and  $\hat{g} = \frac{1}{n} \sum_{i=1}^n \left[ g(X_i, \hat{\theta}) - M' \hat{\nu} \right]$ . Define  $\hat{u}(\hat{\theta}, \hat{\nu}) = n^{-\zeta} \frac{\hat{g}}{\|\hat{g}\|}$ . (Recall the definition of  $u(\theta, \nu)$  in the proof of consistency.) Approximation  $G_n^*(\theta, \nu, \lambda)$  with respect to  $\lambda$  around  $\lambda = 0$  at  $(\theta, \nu, \lambda) = (\hat{\theta}, \hat{\nu}, \hat{u}(\hat{\theta}, \hat{\nu}))$ . Then,

$$\begin{aligned}
&G_n^*(\hat{\theta}, \hat{\nu}, \hat{u}(\hat{\theta}, \hat{\nu})) \\
&= G_n^*(\hat{\theta}, \hat{\nu}, 0) + \frac{\partial G_n^*(\hat{\theta}, \hat{\nu}, 0)}{\partial \lambda'} \hat{u}(\hat{\theta}, \hat{\nu}) + \frac{1}{2} \hat{u}'(\hat{\theta}, \hat{\nu}) \frac{\partial^2 G_n^*(\hat{\theta}, \hat{\nu}, \tilde{\lambda})}{\partial \lambda \partial \lambda'} \hat{u}(\hat{\theta}, \hat{\nu}) \\
&= \hat{g}' \hat{u}(\hat{\theta}, \hat{\nu}) - \frac{1}{2} \hat{u}'(\hat{\theta}, \hat{\nu}) \left( \frac{1}{n} \sum_{i=1}^n \frac{\hat{g}_i \hat{g}_i'}{(1 + \tilde{\lambda}' \hat{g}_i)^2} \right) \hat{u}(\hat{\theta}, \hat{\nu}),
\end{aligned}$$

where  $\tilde{\lambda}$  is located between 0 and  $\hat{u}(\hat{\theta}, \hat{\nu})$ .

Notice that  $\max_{1 \leq i \leq n} |\hat{u}'(\hat{\theta}, \hat{\nu}) \hat{g}_i| \rightarrow_p 0$  and  $\hat{u}(\hat{\theta}, \hat{\nu}) \in \hat{\Lambda}_n(\hat{\theta})$  by Lemma A.1 w.p.a.1. Also, under Assumption 1  $\left\| \frac{1}{n} \sum_{i=1}^n \hat{g}_i \hat{g}_i' \right\| \leq 2 \left( \frac{1}{n} \sum_{i=1}^n \sup_{\theta \in \Theta} \|g(X_i, \theta)\|^2 + K \right) = O_p(1)$ .

Then, w.p.a.1, for some constant  $C$ ,

$$\begin{aligned}
& \hat{g}'\hat{u}(\hat{\theta}, \hat{\nu}) - \frac{1}{2}\hat{u}'(\hat{\theta}, \hat{\nu}) \left( \frac{1}{n} \sum_{i=1}^n \frac{\hat{g}_i \hat{g}'_i}{(1 + \ddot{\lambda}'\hat{g}_i)^2} \right) \hat{u}(\hat{\theta}, \hat{\nu}) \\
&= n^{-\zeta} \|\hat{g}\| - \frac{1}{2}\hat{u}'(\hat{\theta}, \hat{\nu}) \left( \frac{1}{n} \sum_{i=1}^n \frac{\hat{g}_i \hat{g}'_i}{(1 + \ddot{\lambda}'\hat{g}_i)^2} \right) \hat{u}(\hat{\theta}, \hat{\nu}) \\
&\geq n^{-\zeta} \|\hat{g}\| - \frac{1}{2} \max_{1 \leq i \leq n} \left( \frac{1}{(1 + \ddot{\lambda}'\hat{g}_i)^2} \right) \hat{u}'(\hat{\theta}, \hat{\nu}) \left( \frac{1}{n} \sum_{i=1}^n \hat{g}_i \hat{g}'_i \right) \hat{u}(\hat{\theta}, \hat{\nu}) \\
&\geq n^{-\zeta} \|\hat{g}\| - Cn^{-2\zeta}.
\end{aligned} \tag{13}$$

Then,

$$n^{-\zeta} \|\hat{g}\| - Cn^{-2\zeta} \leq G_n^*(\hat{\theta}, \hat{\nu}, \hat{u}(\hat{\theta}, \hat{\nu})) \leq G_n^*(\hat{\theta}, \hat{\nu}, \hat{\lambda}) \leq \sup_{\lambda \in \hat{\Lambda}_n(\theta_{n,0})} G_n^*(\theta_{n,0}, \nu_{n,0}, \lambda) \leq O_p\left(\frac{1}{n}\right), \tag{14}$$

where the first inequality is from (13), the second and third inequalities hold because  $(\hat{\theta}, \hat{\nu}, \hat{\lambda})$  is a saddle point, and the last inequality is from Lemma A.2 with

$$\frac{1}{\sqrt{n}} \sum_{i=1}^n [g(X_i, \theta_{n,0}) - M'\nu_{n,0}] = O_p(1)$$

by Assumption 1(g). Also, by  $\zeta < \frac{1}{2}$ ,  $\zeta - 1 < -\frac{1}{2} < -\zeta$ . Solving (14) for  $\|\hat{g}\|$  gives

$$\|\hat{g}\| \leq O_p(n^{-\zeta}). \tag{15}$$

For a given sequence  $\varepsilon_n \rightarrow 0$ , let  $\bar{\lambda} = \varepsilon_n \hat{g}$ . According to (15)  $\bar{\lambda} = o_p(n^{-\zeta})$ . Hence,  $\bar{\lambda} \in \Lambda_n^\zeta$  w.p.a.1. Then, as in (14), we have

$$\bar{\lambda}'\hat{g} - C\|\bar{\lambda}\|^2 = \varepsilon_n \|\hat{g}\|^2 - C\varepsilon_n^2 \|\hat{g}\|^2 \leq \varepsilon_n \|\hat{g}\|^2 (1 - C\varepsilon_n) \leq O_p\left(\frac{1}{n}\right).$$

For large enough  $n$  the term  $1 - C\varepsilon_n$  is bounded away from zero and it follows that  $\varepsilon_n \|\hat{g}\|^2 = O_p\left(\frac{1}{n}\right)$ . Since  $\varepsilon_n$  is an arbitrary sequence that tends to zero, we deduce that

$$\|\hat{g}\| = O_p\left(\frac{1}{\sqrt{n}}\right),$$

as required. ■

**Lemma 4** Suppose that Assumption 1 is satisfied. Let  $g_n(x, \theta) = I\{x \in \mathcal{X}_n\}g(x, \theta)$  where

$$\mathcal{X}_n = \left\{ x : \sup_{\theta \in \Theta} \|g(x, \theta)\| \leq n^\xi \right\},$$

where  $\frac{1}{\alpha^2} < \xi < \frac{1}{2\alpha}$  and  $\alpha > 2$  as in Assumption 1(i). Define

$$\begin{aligned} q_n(X_i, \theta, \nu) &= \ln [1 + n^{-\zeta} u'(\theta, \nu) g_n(X_i, \theta)] - n^{-\zeta} u'(\theta, \nu) M' \nu \\ \tilde{q}_n(X_i, \theta, \nu) &= \ln [1 + n^{-\zeta} u'(\theta, \nu) g(X_i, \theta)] - n^{-\zeta} u'(\theta, \nu) M' \nu \end{aligned}$$

and assume that  $\|u(\theta, \nu)\| \leq 1$ . Then,

$$\sup_{\theta \in \Theta, \nu \geq 0} \left| \frac{1}{n} \sum_{i=1}^n \left( q_n(X_i, \theta, \nu) - \tilde{q}_n(X_i, \theta, \nu) \right) \right| = o_p(n^{-\zeta}).$$

**Proof of Lemma 4:** By the mean value theorem,

$$\begin{aligned} & \sup_{\theta \in \Theta, \nu \geq 0} \left| \frac{1}{n} \sum_{i=1}^n \{q_n(X_i, \theta, \nu) - \tilde{q}_n(X_i, \theta, \nu)\} \right| \\ &= \sup_{\theta \in \Theta, \nu \geq 0} \left| \frac{1}{n} \sum_{i=1}^n \left( \frac{n^{-\zeta} u'(\theta, \nu) g(X_i, \theta)}{1 + n^{-\zeta} u'_*(\theta, \nu) g(X_i, \theta)} \right) I\{X_i \notin \mathcal{X}_n\} \right| \quad (16) \\ &\leq \max_{1 \leq i \leq n} \sup_{\theta \in \Theta, \nu \geq 0} \left| \frac{n^{-\zeta} u'(\theta, \nu) g(X_i, \theta)}{1 + n^{-\zeta} u'_*(\theta, \nu) g(X_i, \theta)} \right| \frac{1}{n} \sum_{i=1}^n I\left\{ \sup_{\theta \in \Theta} \|g(X_i, \theta)\| > n^\xi \right\} \\ &\leq \frac{1}{n^{\alpha\xi}} \left( \max_{1 \leq i \leq n} \sup_{\theta \in \Theta, \nu \geq 0} \left| \frac{n^{-\zeta} u'(\theta, \nu) g(X_i, \theta)}{1 + n^{-\zeta} u'_*(\theta, \nu) g(X_i, \theta)} \right| \right) \left( \frac{1}{n} \sum_{i=1}^n \sup_{\theta \in \Theta} \|g(X_i, \theta)\|^\alpha \right) \end{aligned}$$

where  $u_*(\theta, \nu)$  is located between 0 and  $u(\theta, \nu)$ . The second term on the right-hand side of (16) can be bounded as follows. According to (1)

$$n^{-\zeta} \max_{1 \leq i \leq n} \sup_{\theta \in \Theta} \|g(X_i, \theta)\| = n^{-\zeta+1/\alpha} O_p(1).$$

Moreover,  $\|u(\theta, \nu)\| \leq 1$ . Therefore,

$$\begin{aligned} \max_{1 \leq i \leq n} \sup_{\theta \in \Theta, \nu \geq 0} \left| \frac{n^{-\zeta} u'(\theta, \nu) g(X_i, \theta)}{1 + n^{-\zeta} u'_*(\theta, \nu) g(X_i, \theta)} \right| &\leq \frac{2n^{-\zeta} \max_{1 \leq i \leq n} \sup_{\theta \in \Theta} \|g(X_i, \theta)\|}{1 - 2n^{-\zeta} \max_{1 \leq i \leq n} \sup_{\theta \in \Theta} \|g(X_i, \theta)\|} \\ &= \frac{n^{-\zeta+1/\alpha} O_p(1)}{1 - n^{-\zeta+1/\alpha} O_p(1)} = n^{-\zeta+1/\alpha} O_p(1). \end{aligned}$$

By Assumption 1(i) and the Markov inequality, the third term on the right-hand side of (16) is  $O_p(1)$ . Since  $\frac{1}{\alpha^2} < \xi < \frac{1}{2\alpha}$ , we are able to deduce that

$$n^\zeta \sup_{\theta \in \Theta, \nu \geq 0} \left| \frac{1}{n} \sum_{i=1}^n \left( q_n(X_i, \theta, \nu) - \tilde{q}_n(X_i, \theta, \nu) \right) \right| = n^{-\alpha\xi + \frac{1}{\alpha}} O_p(1) = o_p(1),$$

as required. ■

## 2 Limit Distributions

Let  $\beta = [\theta', \nu', \lambda']'$ ,  $\beta_{n,0} = [\theta'_{n,0}, \nu'_{n,0}, 0_{1 \times h}]'$ , and abbreviate  $G_n^*(\theta, \nu, \lambda)$  as  $G_n^*(\beta)$ . The objective function is expanded around  $\beta_{n,0}$  as follows:

$$G_n^*(\beta) = G_{nq}^*(\beta) + \frac{1}{n} \mathcal{R}_n(\beta), \quad (17)$$

where

$$G_{nq}^*(\beta) = G_n^*(\beta_{n,0}) + G_n^{*(1)}(\beta_{n,0})(\beta - \beta_{n,0}) + \frac{1}{2}(\beta - \beta_{n,0})' G_n^{*(2)}(\beta_{n,0})(\beta - \beta_{n,0}).$$

We begin by deriving the coefficient matrices for the quadratic approximation of the objective function

$$G_{nq}^*(\beta) = G_n^*(\beta_{n,0}) + G_n^{*(1)}(\beta_{n,0})'(\beta - \beta_{n,0}) + \frac{1}{2}(\beta - \beta_{n,0})' G_n^{*(2)}(\beta_{n,0})(\beta - \beta_{n,0}). \quad (18)$$

A direct calculation shows that

$$G_n^{*(1)}(\beta) = \left[ G_n^{*(1)}(\beta)'_{\theta}, G_n^{*(1)}(\beta)'_{\nu}, G_n^{*(1)}(\beta)'_{\lambda} \right]', \quad (19)$$

where

$$\begin{aligned} G_n^{*(1)}(\beta)_{\theta} &= \frac{1}{n} \sum_{i=1}^n \left( \frac{g^{(1)}(X_i, \theta) \lambda}{1 + \lambda' g(X_i, \theta)} \right), \\ G_n^{*(1)}(\beta)_{\nu} &= -M\lambda, \\ G_n^{*(1)}(\beta)_{\lambda} &= \frac{1}{n} \sum_{i=1}^n \left( \frac{g(X_i, \theta)}{1 + \lambda' g(X_i, \theta)} \right) - M'v. \end{aligned}$$

At  $\beta_{n,0}$  the first derivatives simplify to

$$G_n^{*(1)}(\beta_{n,0}) = [0, 0, n^{-1/2} Z_n']. \quad (20)$$

We proceed by partitioning the matrix of second derivative as follows

$$G_n^{*(2)}(\beta) = \begin{pmatrix} G_n^{*(2)}(\beta)_{\theta\theta'} & G_n^{*(2)}(\beta)_{\theta\nu'} & G_n^{*(2)}(\beta)_{\theta\lambda'} \\ G_n^{*(2)}(\beta)_{\nu\theta'} & G_n^{*(2)}(\beta)_{\nu\nu'} & G_n^{*(2)}(\beta)_{\nu\lambda'} \\ G_n^{*(2)}(\beta)_{\lambda\theta'} & G_n^{*(2)}(\beta)_{\lambda\nu'} & G_n^{*(2)}(\beta)_{\lambda\lambda'} \end{pmatrix}, \quad (21)$$

where

$$\begin{aligned} G_n^{*(2)}(\beta)_{\theta\theta'} &= \frac{1}{n} \sum_{i=1}^n \left( \frac{\sum_{j=1}^h \lambda_j g_j^{(2)}(X_i, \theta)}{1 + \lambda' g(X_i, \theta)} - \frac{g^{(1)}(X_i, \theta) \lambda \lambda' g^{(1)}(X_i, \theta)'}{(1 + \lambda' g(X_i, \theta))^2} \right), \\ G_n^{*(2)}(\beta)_{\theta\nu'} &= 0, \quad G_n^{*(2)}(\beta)_{\nu\nu'} = 0, \quad G_n^{*(2)}(\beta)_{\lambda\nu'} = -M', \\ G_n^{*(2)}(\beta)_{\lambda\theta'} &= \frac{1}{n} \sum_{i=1}^n \left( \frac{g^{(1)}(X_i, \theta)'}{1 + \lambda' g(X_i, \theta)} - \frac{g(X_i, \theta) \lambda' g^{(1)}(X_i, \theta)'}{(1 + \lambda' g(X_i, \theta))^2} \right), \\ G_n^{*(2)}(\beta)_{\lambda\lambda'} &= -\frac{1}{n} \sum_{i=1}^n \frac{g(X_i, \theta) g(X_i, \theta)'}{(1 + \lambda' g(X_i, \theta))^2}. \end{aligned}$$

At  $\beta_{n,0}$  the second derivatives simplify to

$$G_n^{*(2)}(\beta_{n,0}) = \begin{bmatrix} 0 & 0 & Q_n \\ 0 & 0 & -M \\ Q_n' & -M' & -J_n \end{bmatrix}. \quad (22)$$

The objective function  $G_{nq}^*(\beta)$  in terms of the transformed parameters is:

$$\begin{aligned} \mathcal{G}_{nq}^*(\phi, l) &= -\frac{1}{2}(l - J_n^{-1}[Z_n - R'_n(\phi - \phi_{n,0})])' J_n (l - J_n^{-1}[Z_n - R'_n(\phi - \phi_{n,0})]) \\ &\quad + \frac{1}{2}(Z_n - R'_n(\phi - \phi_{n,0}))' J_n^{-1} (Z_n - R'_n(\phi - \phi_{n,0})) \end{aligned} \quad (23)$$

The function  $\mathcal{G}_{nq}^*(\phi, l)$  is maximized with respect to  $l \in \mathbb{R}^h$  by

$$\tilde{l}_q(\phi) = J_n^{-1} (Z_n - R'_n(\phi - \phi_{n,0})) \quad (24)$$

and the concentrated objective function is:

$$\bar{\mathcal{G}}_{nq}^*(\phi) = \mathcal{G}_{nq}^*(\phi, \tilde{l}_q(\phi)) = \frac{1}{2}(Z_n - R'_n(\phi - \phi_{n,0}))' J_n^{-1} (Z_n - R'_n(\phi - \phi_{n,0})). \quad (25)$$

## 2.1 Assumptions

**Assumption 2** (a) The true parameter  $\theta_0$  exists in an interior of  $\Theta$ ;

(b)  $g(X_i, \theta)$  is twice continuously differentiable with respect to  $\theta$ ;

(c) the minimum eigenvalue of  $(\mathbb{E}_n[g_1^{(1)}(X_i, \theta)])(\mathbb{E}_n[g_1^{(1)}(X_i, \theta)])'$  is bounded below by a constant  $K > 0$ ;

(d)  $\mathbb{E}_n[\sup_{\theta \in \Theta} \|g^{(1)}(X_i, \theta)\|^2] \leq K < \infty$ ,  $\mathbb{E}_n[\sup_{\theta \in \Theta} \|g_j^{(2)}(X_i, \theta)\|] \leq K < \infty$  for  $j = 1, \dots, h$ ;

(e) for any  $\theta$  and  $\theta^*$ ,  $\|g_j^{(2)}(X_i, \theta) - g_j^{(2)}(X_i, \theta^*)\| \leq L_j(X_i) l_j(\|\theta - \theta^*\|)$ , for some measurable function  $L_j$  of  $X_i$  such that  $\sup_n \mathbb{E}_n(L_j(X_i)) < \infty$ , and  $l_j(y) \downarrow 0$  as  $y \downarrow 0$ .

**Assumption 3** (a) For each  $\theta$ ,  $Q_n(\theta) \xrightarrow{p} Q(\theta)$  and  $J_n(\theta) \xrightarrow{p} J(\theta)$ . (b) For each  $\theta$ ,  $\frac{1}{n} \sum_{i=1}^n g_j^{(2)}(X_i, \theta) \xrightarrow{p} (\lim_{n \rightarrow \infty} \mathbb{E}_n[g_j^{(2)}(X_i, \theta)])$ . (c)  $Z_n \implies Z$ , where  $Z \sim \mathcal{N}(0, J - M' \nu_0 \nu_0' M)$ .

## 2.2 Negligible Remainder

**Lemma 5** Suppose Assumptions 1 to 1 are satisfied, then for all  $\gamma_n \rightarrow 0$

$$\sup_{\beta \in \mathcal{B}_n: \|\beta - \beta_{n,0}\| \leq \gamma_n} \frac{|\mathcal{R}_n(\beta)|}{(1 + \|\sqrt{n}(\beta - \beta_{n,0})\|^2)} = o_p(1), \quad (26)$$

where  $\mathcal{R}_n(\beta)$  is the remainder term in (17).

**Proof of Lemma 5:** By Lemma 1(a) of Andrews (1999), it is sufficient to prove

$$\sup_{\beta \in \mathcal{B}_n: \|\beta - \beta_{n,0}\| \leq \gamma_n} \left\| G_n^{*(2)}(\beta) - G_n^{*(2)}(\beta_{n,0}) \right\| = o_p(1),$$

for every sequence  $\gamma_n \rightarrow 0$ .  $G_n^{*(2)}$  is defined in (21). To verify this sufficient condition we will subsequently show that

- (i)  $\sup_{\beta \in \mathcal{B}_n: \|\beta - \beta_{n,0}\| \leq \gamma_n} \left\| G_n^{*(2)}(\beta)_{\theta\theta'} - G_n^{*(2)}(\beta_{n,0})_{\theta\theta'} \right\| = o_p(1),$
- (ii)  $\sup_{\beta \in \mathcal{B}_n: \|\beta - \beta_{n,0}\| \leq \gamma_n} \left\| G_n^{*(2)}(\beta)_{\lambda\theta'} - G_n^{*(2)}(\beta_{n,0})_{\lambda\theta'} \right\| = o_p(1),$
- (iii)  $\sup_{\beta \in \mathcal{B}_n: \|\beta - \beta_{n,0}\| \leq \gamma_n} \left\| G_n^{*(2)}(\beta)_{\lambda\lambda'} - G_n^{*(2)}(\beta_{n,0})_{\lambda\lambda'} \right\| = o_p(1).$

We begin by showing that

$$\sup_{\beta \in \mathcal{B}_n} \left| \frac{1}{1 + \lambda'g(X_i, \theta)} \right| = O_p(1). \quad (27)$$

For any given  $0 < \delta < \frac{1}{2}$ , set  $K = \frac{1}{1-\delta}$ . Then, since  $\sup_{1 \leq i \leq n, \beta \in \mathcal{B}_n} |\lambda'g(X_i, \theta)| \leq \delta$  implies  $\sup_{1 \leq i \leq n, \beta \in \mathcal{B}_n} \left| \frac{1}{1 + \lambda'g(X_i, \theta)} \right| \leq K$ ,

$$P \left\{ \sup_{1 \leq i \leq n, \beta \in \mathcal{B}_n} \left| \frac{1}{1 + \lambda'g(X_i, \theta)} \right| > K \right\} \leq P \left\{ \sup_{1 \leq i \leq n, \beta \in \mathcal{B}_n} |\lambda'g(X_i, \theta)| > \delta \right\} \longrightarrow 0,$$

which proves (27). The convergence result for the upper bound can be deduced from Lemma 1.

(i) Notice that

$$\begin{aligned} & \sup_{\beta \in \mathcal{B}_n: \|\beta - \beta_{n,0}\| \leq \gamma_n} \left\| \frac{1}{n} \sum_{i=1}^n \left( \frac{\lambda_j g_j^{(2)}(X_i, \theta)}{1 + \lambda'g(X_i, \theta)} \right) \right\| \\ & \leq \sup_{\lambda \in \Lambda_n^\zeta} |\lambda_j| \left( \sup_{\beta \in \mathcal{B}_n, 1 \leq i \leq n} \left| \frac{1}{1 + \lambda'g(X_i, \theta)} \right| \right) \left( \sup_{\theta \in \Theta} \frac{1}{n} \sum_{i=1}^n \|g_j^{(2)}(X_i, \theta)\| \right) \\ & = O(n^{-\zeta}) O_p(1) O_p(1) = o_p(1), \end{aligned}$$

where the last inequality holds by the definition of  $\Lambda_n^\zeta$ , (27) and (5). Moreover,

$$\begin{aligned} & \sup_{\beta \in \mathcal{B}_n: \|\beta - \beta_{n,0}\| \leq \gamma_n} \left\| \frac{1}{n} \sum_{i=1}^n \left( \frac{g^{(1)}(X_i, \theta)' \lambda \lambda' g^{(1)}(X_i, \theta)}{(1 + \lambda'g(X_i, \theta))^2} \right) \right\| \\ & \leq \sup_{\lambda \in \Lambda_n^\zeta} \|\lambda\|^2 \left( \sup_{\beta \in \mathcal{B}_n, 1 \leq i \leq n} \frac{1}{(1 + \lambda'g(X_i, \theta))^2} \right) \left( \sup_{\theta \in \Theta} \frac{1}{n} \sum_{i=1}^n \|g^{(1)}(X_i, \theta)\| \right) \\ & = O(n^{-2\zeta}) O_p(1) O_p(1) = o_p(1). \end{aligned}$$

The last inequality holds by the definition of  $\Lambda_n^\zeta$ , (27) and (5).

(ii) Apply the triangle inequality to

$$\begin{aligned}
 & \sup_{\beta \in \mathcal{B}_n: \|\beta - \beta_{n,0}\| \leq \gamma_n} \left\| \frac{1}{n} \sum_{i=1}^n \left( \frac{g^{(1)}(X_i, \theta)}{1 + \lambda' g(X_i, \theta)} - g^{(1)}(X_i, \theta_0) \right) \right\| \\
 & \leq \sup_{\beta \in \mathcal{B}_n: \|\beta - \beta_{n,0}\| \leq \gamma_n} \left\| \frac{1}{n} \sum_{i=1}^n \left( \frac{g^{(1)}(X_i, \theta)}{1 + \lambda' g(X_i, \theta)} - g^{(1)}(X_i, \theta) \right) \right\| \\
 & \quad + \sup_{\theta \in \Theta} \left\| \frac{1}{n} \sum_{i=1}^n \left( g^{(1)}(X_i, \theta) - \mathbb{E}_n \left[ g^{(1)}(X_i, \theta) \right] \right) \right\| \\
 & \quad + \sup_{\theta \in \Theta: \|\theta - \theta_0\| \leq \gamma_n} \left\| \mathbb{E}_n \left[ g^{(1)}(X_i, \theta) \right] - \mathbb{E}_n \left[ g^{(1)}(X_i, \theta_0) \right] \right\| \\
 & \quad + \left\| \frac{1}{n} \sum_{i=1}^n \left( g^{(1)}(X_i, \theta_0) - \mathbb{E}_n \left[ g^{(1)}(X_i, \theta_0) \right] \right) \right\| \\
 & = I_d + o_p(1) + o_p(1) + o_p(1),
 \end{aligned}$$

where the last equality holds by (5) and (4). Next,

$$\begin{aligned}
 I_d & \leq \sup_{\beta \in \mathcal{B}_n} |\lambda' g(X_i, \theta)| \left( \sup_{\beta \in \mathcal{B}_n} \left| \frac{1}{1 + \lambda' g(X_i, \theta)} \right| \right) \left( \sup_{\theta \in \Theta} \frac{1}{n} \sum_{i=1}^n \left\| g^{(1)}(X_i, \theta) \right\| \right) \\
 & = o_p(1) O_p(1) O_p(1) O_p(1) = o_p(1)
 \end{aligned}$$

by Lemma 1, (27), and (5). Moreover,

$$\begin{aligned}
 & \sup_{\beta \in \mathcal{B}_n: \|\beta - \beta_{n,0}\| \leq \gamma_n} \left\| \frac{1}{n} \sum_{i=1}^n \frac{g(X_i, \theta)}{1 + \lambda' g(X_i, \theta)} \frac{\lambda' g^{(1)}(X_i, \theta)}{1 + \lambda' g(X_i, \theta)} \right\| \\
 & \leq \sup_{\lambda \in \Lambda_n^c} \|\lambda\| \left( \sup_{\beta \in \mathcal{B}_n, 1 \leq i \leq n} \frac{1}{(1 + \lambda' g(X_i, \theta))^2} \right) \left( \frac{1}{n} \sum_{i=1}^n \sup_{\theta \in \Theta} \|g(X_i, \theta)\|^2 \right)^{1/2} \\
 & \quad \times \left( \frac{1}{n} \sum_{i=1}^n \sup_{\theta \in \Theta} \left\| g^{(1)}(X_i, \theta) \right\|^2 \right)^{1/2} \\
 & = O(n^{-\zeta}) O_p(1) O_p(1) = o_p(1).
 \end{aligned}$$

(iii) Similar as before, we have

$$\begin{aligned}
 & \sup_{\beta \in \mathcal{B}_n: \|\beta - \beta_{n,0}\| \leq \gamma_n} \left\| \frac{1}{n} \sum_{i=1}^n \left( \frac{g(X_i, \theta) g(X_i, \theta)'}{(1 + \lambda' g(X_i, \theta))^2} - g(X_i, \theta_0) g(X_i, \theta_0)' \right) \right\| \\
 & \leq \sup_{\beta \in \mathcal{B}_n: \|\beta - \beta_{n,0}\| \leq \gamma_n} \left\| \frac{1}{n} \sum_{i=1}^n \left( \frac{g(X_i, \theta) g(X_i, \theta)'}{(1 + \lambda' g(X_i, \theta))^2} - g(X_i, \theta) g(X_i, \theta)' \right) \right\| \\
 & \quad + \sup_{\Theta} \left\| \frac{1}{n} \sum_{i=1}^n (g(X_i, \theta) g(X_i, \theta)' - \mathbb{E}_n [g(X_i, \theta) g(X_i, \theta)']) \right\| \\
 & \quad + \sup_{\Theta} \left\| \mathbb{E}_n [g(X_i, \theta) g(X_i, \theta)'] - \mathbb{E}_n [g(X_i, \theta_0) g(X_i, \theta_0)'] \right\| \\
 & \quad + \sup_{\Theta} \left\| \frac{1}{n} \sum_{i=1}^n (g(X_i, \theta_0) g(X_i, \theta_0)' - \mathbb{E}_n [g(X_i, \theta_0) g(X_i, \theta_0)']) \right\| \\
 & = \sup_{\beta \in \mathcal{B}_n: \|\beta - \beta_{n,0}\| \leq \gamma_n} \left\| \frac{1}{n} \sum_{i=1}^n \left( \frac{g(X_i, \theta) g(X_i, \theta)'}{(1 + \lambda' g(X_i, \theta))^2} - g(X_i, \theta) g(X_i, \theta)' \right) \right\| + o_p(1).
 \end{aligned}$$

Next,

$$\begin{aligned}
 & \sup_{\beta \in \mathcal{B}_n: \|\beta - \beta_{n,0}\| \leq \gamma_n} \left\| \frac{1}{n} \sum_{i=1}^n \left( \frac{g(X_i, \theta) g(X_i, \theta)'}{(1 + \lambda' g(X_i, \theta))^2} - g(X_i, \theta) g(X_i, \theta)' \right) \right\| \\
 & \leq \sup_{\beta \in \mathcal{B}_n, 1 \leq i \leq n} |\lambda' g(X_i, \theta)| \left( \sup_{\beta \in \mathcal{B}_n, 1 \leq i \leq n} \frac{1}{|1 + \lambda' g(X_i, \theta)|} \right) \\
 & \quad \times \left( \sup_{\beta \in \mathcal{B}_n, 1 \leq i \leq n} \frac{1}{|1 + \lambda' g(X_i, \theta)|} + 1 \right) \left( \sup_{\theta \in \Theta} \frac{1}{n} \sum_{i=1}^n \|g(X_i, \theta)\|^2 \right) \\
 & = o_p(1) O_p(1) O_p(1) O_p(1) = o_p(1). \quad \blacksquare
 \end{aligned}$$

### 2.3 $\sqrt{n}$ Consistency

**Theorem 2** *Suppose Assumptions 1 – 3 are satisfied. Then, (i)  $\sqrt{n}(\tilde{\beta}_{nq} - \beta_{n,0}) = O_p(1)$ , (ii)  $\sqrt{n}(\hat{\beta}_n - \beta_{n,0}) = O_p(1)$ , (iii)  $nG_n^*(\hat{\beta}_n) = nG_{nq}^*(\hat{\beta}_n) + o_p(1)$ , (iv)  $nG_{nq}^*(\hat{\beta}_n) = nG_{nq}^*(\tilde{\beta}_{nq}) + o_p(1)$ , and (v)  $nG_n^*(\hat{\beta}_n) = nG_{nq}^*(\tilde{\beta}_{nq}) + o_p(1)$ .*

Theorem 2 establishes that  $\hat{\beta}_n$  and  $\tilde{\beta}_{nq}$  are  $\sqrt{n}$ -consistent. Moreover, the theorem states that the discrepancy between  $G_n^*(\beta)$  evaluated at  $\hat{\beta}_n$  and  $G_{nq}^*(\beta)$  evaluated at  $\tilde{\beta}_{nq}$  vanishes. Thus, the large-sample behavior of likelihood ratios can be approximated by the behavior of  $G_{nq}^*(\tilde{\beta}_{nq})$ . Let

$$\hat{b} = [\hat{\phi}', \hat{l}'(\hat{\phi})]' \quad \text{and} \quad \tilde{b}_q = [\tilde{\phi}'_q, \tilde{l}'_q(\tilde{\phi}_q)]'$$

be re-scaled versions of  $\hat{\beta}_n$  and  $\tilde{\beta}_{nq}$ . To prove the theorem, we will introduce a third estimator

$$\hat{b}_q = [\hat{\phi}'_q, \hat{l}'_q(\hat{\phi}_q)]'$$

where

$$\hat{l}_q(\phi) = \operatorname{argmax}_{l \in L_n(\phi)} \mathcal{G}_{nq}^*(\phi, l), \quad \hat{\phi}_q = \operatorname{argmin}_{\phi \in \Phi_n} \mathcal{G}_{nq}^*(\phi, \hat{l}_q(\phi)).$$

$\hat{b}_q$  is based on the quadratic approximation of the objective function, but the domains of  $\phi$  and  $l$  are restricted. In slight abuse of notation we let  $B_n = \Phi_n(u_0) \otimes L_n$ .

**Proof of Theorem 2:** (i) Follows from Lemma 7.

(ii) According to Lemma 2,  $\hat{\lambda}(\hat{\theta}, \hat{\nu}) = O_p(n^{-1/2})$ . It remains to show that

$$\hat{\phi} = \left[ \sqrt{n}(\hat{\theta} - \theta_{n,0})', u'_{n,0} + \sqrt{n}(\hat{\nu} - \nu_{n,0})' \right]'$$

is stochastically bounded. The saddlepoint property implies that

$$0 = \mathcal{G}_n^*(\hat{\phi}, 0) \leq \mathcal{G}_n^*(\hat{\phi}, \hat{l}(\hat{\phi})) \leq \mathcal{G}_n^*(0, \hat{l}(0)). \quad (28)$$

Then using the quadratic approximation (17), the bound for the remainder term given in Lemma 5 and the definition of  $\hat{l}$  and  $\hat{\phi}$  we obtain

$$\begin{aligned} \mathcal{G}_n^*(\hat{\phi}, \hat{l}(\hat{\phi})) &= \mathcal{G}_{nq}^*(\hat{\phi}, \hat{l}(\hat{\phi})) + (1 + \|\hat{\phi} - \phi_{n,0}\|^2 + \|\hat{l}(\hat{\phi})\|^2) o_p(1) \\ &= \frac{1}{2} (Z_n - R'_n(\hat{\phi} - \phi_{n,0}))' J_n^{-1} (Z_n - R'_n(\hat{\phi} - \phi_{n,0})) \\ &\quad - \frac{1}{2} (\hat{l}(\hat{\phi}) - J_n^{-1} [Z_n - R'_n(\hat{\phi} - \phi_{n,0})])' J_n (\hat{l}(\hat{\phi}) - J_n^{-1} [Z_n - R'_n(\hat{\phi} - \phi_{n,0})]) \\ &\quad + (1 + \|\hat{\phi} - \phi_{n,0}\|^2 + \|\hat{l}(\hat{\phi})\|^2) o_p(1) \\ &= \frac{1}{2} (Z_n - R'_n(\hat{\phi} - \phi_{n,0}))' J_n^{-1} (Z_n - R'_n(\hat{\phi} - \phi_{n,0})) + (1 + \|\hat{\phi} - \phi_{n,0}\|^2 + \|\hat{l}(\hat{\phi})\|^2) o_p(1), \end{aligned} \quad (29)$$

where  $\phi_{n,0} = [0, u'_{n,0}]'$ . The last equality is a consequence of Lemma 8. Similarly, we can deduce from Lemmas 2, 5, and Assumptions 1 and 3 that

$$\mathcal{G}_n^*(0, \hat{l}(0)) = -\frac{1}{2} \hat{l}(0)' J_n \hat{l}(0) + Z_n' \hat{l}(0) + (1 + \|\hat{l}(0)\|^2) o_p(1) = O_p(1). \quad (30)$$

Hence, from (28), (29), and (30) we obtain the inequality

$$0 \leq \frac{1}{2} (Z_n + o_p(1) - R'_n(\hat{\phi} - \phi_{n,0}))' J_n^{-1} (Z_n + o_p(1) - R'_n(\hat{\phi} - \phi_{n,0})) \leq O_p(1). \quad (31)$$

Notice that  $Z_n + o_p(1) = O_p(1)$ . According to Assumption 1,  $R_n$  is full rank and  $J_n$  is positive definite w.p.a. 1. Therefore, (31) implies that  $\hat{\phi} - \phi_{n,0}$  is stochastically bounded.

(iii) We deduce from Lemma 5 and Part (ii) that

$$\begin{aligned} nG_{nq}^*(\hat{\beta}_n) &= \mathcal{G}_{nq}^*(\sqrt{n}(\hat{\beta}_n - \beta_{n,0})) + (1 + \|\sqrt{n}(\hat{\beta}_n - \beta_{n,0})\|^2) o_p(1) \\ &= nG_{nq}^*(\hat{\beta}_n) + O_p(1) o_p(1). \end{aligned}$$

(iv) We proceed by establishing  $o_p(1)$  bounds for  $nG_{nq}^*(\hat{\beta}_n) - nG_{nq}^*(\tilde{\beta}_{nq})$ .

We begin with the upper bound. Using (iii) we can rewrite the differential as

$$\begin{aligned} nG_{nq}^*(\hat{\beta}_n) - nG_{nq}^*(\tilde{\beta}_{nq}) &= \mathcal{G}_n^*(\hat{\phi}, \hat{l}(\hat{\phi})) + o_p(1) - \mathcal{G}_{nq}^*(\tilde{\phi}_q, \tilde{l}_q(\tilde{\phi}_q)) \\ &\leq \mathcal{G}_n^*(\hat{\phi}_q, \hat{l}(\hat{\phi}_q)) - \mathcal{G}_{nq}^*(\tilde{\phi}_q, \tilde{l}(\tilde{\phi}_q)) + o_p(1). \end{aligned} \quad (32)$$

Replacing  $\hat{\phi}$  by  $\hat{\phi}_q$  raises  $\mathcal{G}_n^*$ , whereas substituting  $\tilde{l}_q$  with  $\hat{l}$  lowers  $\mathcal{G}_{nq}^*$ . Using Lemma 5 the first term on the right-hand side of (32) can be rewritten as

$$\begin{aligned} \mathcal{G}_n^*(\hat{\phi}_q, \hat{l}(\hat{\phi}_q)) &= \mathcal{G}_{nq}^*(\hat{\phi}_q, \hat{l}(\hat{\phi}_q)) + o_p(1) \left(1 + \|\hat{\phi}_q - \phi_{n,0}\|^2 + \|\hat{l}(\hat{\phi}_q)\|^2\right) \\ &= \mathcal{G}_{nq}^*(\hat{\phi}_q, \hat{l}(\hat{\phi}_q)) + o_p(1). \end{aligned} \quad (33)$$

The second equality in (33) is a consequence of Lemmas 2 and 7. According to Lemma 8

$$\hat{l}(\bar{\phi}) = (J_n + o_p(1))^{-1} [Z_n - (R'_n + o_p(1))(\bar{\phi} - \phi_{n,0})]$$

for  $\bar{\phi} = O_p(1)$ . Hence,

$$\hat{l}(\tilde{\phi}_q) - \hat{l}(\hat{\phi}_q) = -(J_n + o_p(1))^{-1} [(R'_n + o_p(1))(\tilde{\phi}_q - \hat{\phi}_q)] = o_p(1)$$

by Lemma 7. Since  $\mathcal{G}_{nq}^*(\phi, l)$  is continuous in its arguments we can now express the second term on the right-hand side of (32) as

$$\mathcal{G}_{nq}^*(\tilde{\phi}_q, \hat{l}(\tilde{\phi}_q)) = \mathcal{G}_{nq}^*(\hat{\phi}_q, \hat{l}(\hat{\phi}_q)) + o_p(1) \quad (34)$$

Plugging (33) and (34) into (32) we obtain the upper bound

$$n\mathcal{G}_{nq}^*(\hat{\beta}_n) - n\mathcal{G}_{nq}^*(\tilde{\beta}_{nq}) \leq o_p(1).$$

Using similar arguments, we can establish a lower bound as follows:

$$\begin{aligned} n\mathcal{G}_{nq}^*(\hat{\beta}_n) - n\mathcal{G}_{nq}^*(\tilde{\beta}_{nq}) &= \mathcal{G}_n^*(\hat{\phi}, \hat{l}(\hat{\phi})) - \mathcal{G}_{nq}^*(\tilde{\phi}_q, \tilde{l}_q(\tilde{\phi}_q)) + o_p(1) \\ &\geq \mathcal{G}_n^*(\hat{\phi}, \hat{l}_q(\hat{\phi})) - \mathcal{G}_{nq}^*(\hat{\phi}, \tilde{l}_q(\hat{\phi})) + o_p(1) \\ &= \mathcal{G}_n^*(\hat{\phi}, \hat{l}_q(\hat{\phi})) - \mathcal{G}_{nq}^*(\hat{\phi}, \hat{l}_q(\hat{\phi})) + o_p(1) \\ &= o_p(1) \end{aligned}$$

which proves (iv). ■

(v) Follows from parts (iii) and (iv).

### 2.3.1 Technical Lemmas

**Lemma 6** *Suppose Assumptions 1 to 3 are satisfied. Then,  $\tilde{b}_q$  exists uniquely w.p.a. 1.*

**Proof of Lemma 6:** The subsequent statements are true w.p.a. 1. Notice that  $\bar{\mathcal{G}}_{nq}^*(\phi)$ , defined in (25), is strictly convex function of  $\phi$  because  $R'_n = [-Q'_n, M']$  is a full rank matrix under Assumption 1(c) and  $J_n^{-1}$  is positive definite. Hence,  $R_n J_n^{-1} R'_n$  is a positive definite matrix. Moreover, the domain  $\Phi$  is convex. Therefore,  $\tilde{\phi}_q$  is unique. Finally, from (24) we deduce that  $\tilde{l}_q$  exists uniquely. ■

**Lemma 7** *Suppose Assumptions 1 to 3 are satisfied. Then*

$$(i) \tilde{b}_q = O_p(1),$$

$$(ii) \hat{b}_q = \tilde{b}_q + o_p(1).$$

**Proof of Lemma 7:**

**Proof of (i):** We will show that  $\tilde{\phi}_q = O_p(1)$ . For notational simplicity, denote

$$A_{1n} = R_n J_n^{-1} R_n', \quad A_{2n} = A_{1n}^{-1} R_n J_n^{-1} Z_n, \quad \text{and} \quad A_{3n} = Z_n' J_n^{-1} Z_n - A_{2n}' A_{1n} A_{2n},$$

and write the concentrated quadratic objective function (25) as

$$\bar{\mathcal{G}}_{nq}^*(\phi) = \frac{1}{2} (\phi - \phi_{n,0} + A_{2n})' A_{1n} (\phi - \phi_{n,0} + A_{2n}) + \frac{1}{2} A_{3n}.$$

Observe that  $J_n$ ,  $R_n$ , and  $Z_n$  converge weakly according to Assumptions 1 and 3. Moreover based on Assumption 1,  $A_{1n}$  is positive definite w.p.a. 1. Let

$$\bar{\phi}_q = \operatorname{argmin}_{\phi \in \mathbb{R}^{m+h_2}} \bar{\mathcal{G}}_{nq}^*(\phi) = \phi_{n,0} - A_{2n} = O_p(1).$$

Notice that  $\tilde{\phi}_q$  is the projection of  $\bar{\phi}_q$  onto the set  $\Phi(u_0)$  with respect to the inner product  $\langle x, y \rangle = x' A_{1n} y$ . Then,

$$\|\tilde{\phi}_q\| \leq \lambda_{\min}^{-1}(A_{1n}) \langle \tilde{\phi}_q, \bar{\phi}_q \rangle^{1/2} \leq \lambda_{\min}^{-1}(A_{1n}) \langle \bar{\phi}_q, \bar{\phi}_q \rangle^{1/2} = O_p(1)$$

where  $\lambda_{\min}(A_{1n})$  denotes the smallest eigenvalue of  $A_{1n}$  and is strictly positive w.p.a. 1. Finally, from (24) we can deduce that  $\tilde{l}_q(\tilde{\phi}_q) = O_p(1)$ .

**Proof of (ii):** According to Lemma 6 the saddlepoint problem  $\min_{\phi \in \Phi(u_0)} \max_{l \in \mathbb{R}^h} \mathcal{G}_{nq}^*(\phi, l)$  has a unique solution  $\tilde{b}_q$  on the domain  $B = \Phi(u_0) \otimes \mathbb{R}^h$ . Since  $B_n \subset B$  for any  $\epsilon > 0$

$$\begin{aligned} P \left\{ \|\hat{b}_q - \tilde{b}_q\| > \epsilon \right\} &\leq P \left\{ \tilde{b}_q \in B \setminus B_n \right\} \\ &\leq P \left\{ \tilde{b}_q \in B \setminus (\Phi_n(u_0) \otimes \sqrt{n} \Lambda_n^\zeta) \right\} + o(1), \end{aligned}$$

where the  $o(1)$  term in the last line holds by Lemma 1(ii). The set  $\sqrt{n} \Lambda_n^\zeta$  consists of the elements in  $\Lambda_n^\zeta$  multiplied by  $\sqrt{n}$  and expands to  $\mathbb{R}^h$  because  $\zeta < 1/2$ . Since  $\theta_0$  is in the interior of  $\Theta$ , the first  $m$  ordinates of  $\Phi_n(u_0)$  expand to  $\mathbb{R}^m$ . Ordinate  $m+j$  expands to  $\mathbb{R}$  if  $u_{0,j} = \infty$  and to  $\mathbb{R}^+$  otherwise. Since  $\tilde{b}_q = O_p(1)$ , we deduce  $P\{\tilde{b}_q \in B \setminus (\Phi_n(u_0) \otimes \sqrt{n} \Lambda_n^\zeta)\} = o(1)$ . Therefore  $\hat{b}_q = \tilde{b}_q + o_p(1)$ , as required. ■

**Lemma 8** *Suppose that Assumptions 1 to 3 are satisfied. Let  $\bar{\theta} \in \Theta$  and  $\bar{\nu} \geq 0$  be sequences such that  $\bar{\theta} - \theta_{n,0} \xrightarrow{p} 0$  and  $\bar{\nu} - \nu_{n,0} \xrightarrow{p} 0$ . Let  $\hat{l}(\bar{\phi}) = \sqrt{n} \hat{\lambda}(\bar{\theta}, \bar{\nu})$ , and  $\bar{\phi} = [\bar{s}', \bar{u}']$ , where  $\bar{s} = \sqrt{n}(\bar{\theta} - \theta_{n,0})$  and  $\bar{u} = u_{n,0} + \sqrt{n}(\bar{\nu} - \nu_{n,0})$ . Then*

$$0 = Z_n - (R_n' + o_p(1))(\bar{\phi} - \phi_{n,0}) - (J_n + o_p(1))\hat{l}(\bar{\phi}).$$

**Proof of Lemma 8:** In view of Lemmas 1(ii) and 2, we deduce that  $\hat{\lambda}(\bar{\theta}, \bar{v})$  is in the interior of  $\hat{\Lambda}(\bar{\theta})$  w.p.a. 1. Hence,  $\hat{\lambda}$  satisfies the first-order conditions associated with  $\max_{\lambda \in \hat{\Lambda}(\bar{\theta})} G_n^*(\bar{\theta}, \bar{v}, \lambda)$ :

$$0 = \frac{1}{n} \sum_{i=1}^n \frac{g(X_i, \bar{\theta})}{1 + \hat{\lambda}' g(X_i, \bar{\theta})} - M' \bar{v}.$$

We now apply the mean-value theorem and multiply by  $\sqrt{n}$ :

$$0 = \sqrt{n} G_n^{*(1)}(\beta_{n,0})_{\lambda} + G_n^{*(2)}(\beta_*)_{\lambda \theta' \bar{s}} - M'(\bar{u} - u_{n,0}) + G_n^{*(2)}(\beta_*)_{\lambda \lambda'} \hat{l},$$

where  $\beta_*$  lies on the line joining  $\beta_{n,0}$  and  $\bar{\beta} = [\bar{\theta}', \bar{v}', \hat{\lambda}(\bar{\theta}, \bar{v})']'$ . The matrices  $G_n^{*(1)}(\beta)$  and  $G_n^{*(2)}(\beta)$  and their partitions are defined in (19) and (21). Using the same arguments as in the proof of Lemma 5 and the definitions of  $J_n$ ,  $Q_n$ ,  $R_n$ , and  $Z_n$  we obtain the desired result. ■