

# Appendix B

## B.1 Proof of Theorem 4.1

Put  $\Delta = \boldsymbol{\eta} - \boldsymbol{\eta}^0$  and  $A_i = (a_{isr})$ , and observe that, by (4.5), for  $\boldsymbol{\eta}$  within any given but fixed radius of  $\boldsymbol{\eta}^0$ ,

$$\begin{aligned}\lambda_s(X_i, \boldsymbol{\eta}) &= \lambda_s(X_i, \boldsymbol{\eta}^0) + \lambda'_s(X_i, \boldsymbol{\eta}^0)^\top \Delta + \Theta_{is}(\boldsymbol{\eta}) \|\boldsymbol{\eta} - \boldsymbol{\eta}^0\|^2, \\ a_{isr}(X_i, \boldsymbol{\eta}) &= a_{isr}(X_i, \boldsymbol{\eta}^0) + a'_{isr}(X_i, \boldsymbol{\eta}^0)^\top \Delta + \Theta_{isr}(\boldsymbol{\eta}) \|\boldsymbol{\eta} - \boldsymbol{\eta}^0\|^2,\end{aligned}\tag{B.1}$$

where  $a'_{isr}(X_i, \boldsymbol{\eta}) = \partial a_{isr}(X_i, \boldsymbol{\eta}) / \partial \boldsymbol{\eta}$  and, here and below,  $\Theta_{i\dots}(\boldsymbol{\eta})$  denotes a generic random variable satisfying, with probability 1,  $|\Theta_{i\dots}(\boldsymbol{\eta})| \leq C_1$  for  $\|\boldsymbol{\eta} - \boldsymbol{\eta}^0\| \leq C_2$  and  $\|X_i\| \leq C_2$ , for any  $C_2 > 0$ , where  $C_1$  depends on  $C_2$  but not on  $i$ . Hence, for distinct integers  $t_1, \dots, t_\ell$  depending on  $s_1, \dots, s_\ell$ ,

$$\begin{aligned}Q_{ir}(\boldsymbol{\eta}) &= \sum_{\ell=1}^r \sum_{s_1, \dots, s_\ell} \nu_{s_1, \dots, s_\ell}(r) \left( \prod_{t=1}^{\ell} W_{is_{t}} \right) \\ &\quad \times \left\{ \lambda_{s_1}(X_i, \boldsymbol{\eta}^0) + \lambda'_{s_1}(X_i, \boldsymbol{\eta}^0)^\top \Delta + \Theta_{it_1}(\boldsymbol{\eta}) \|\boldsymbol{\eta} - \boldsymbol{\eta}^0\|^2 \right\} \\ &\quad \times \dots \times \left\{ \lambda_{s_\ell}(X_i, \boldsymbol{\eta}^0) + \lambda'_{s_\ell}(X_i, \boldsymbol{\eta}^0)^\top \Delta + \Theta_{it_\ell}(\boldsymbol{\eta}) \|\boldsymbol{\eta} - \boldsymbol{\eta}^0\|^2 \right\} \\ &= \sum_{\ell=1}^r \sum_{s_1, \dots, s_\ell} \nu_{s_1, \dots, s_\ell}(r) \left( \prod_{t=1}^{\ell} W_{is_{t}} \right) \left[ \prod_{t=1}^{\ell} \lambda_{s_t}(X_i, \boldsymbol{\eta}^0) \right. \\ &\quad \left. + \sum_{t=1}^{\ell} \left\{ \prod_{u: u \neq t} \lambda_{s_u}(X_i, \boldsymbol{\eta}^0) \right\} \lambda'_{s_t}(X_i, \boldsymbol{\eta}^0)^\top \Delta + \Theta_i(\boldsymbol{\eta}) \|\boldsymbol{\eta} - \boldsymbol{\eta}^0\|^2 \right] \\ &= Q_{ir}(X_i, \boldsymbol{\eta}^0) + u_{ir}^\top \Delta + \Theta_i(\boldsymbol{\eta}) \|\boldsymbol{\eta} - \boldsymbol{\eta}^0\|^2.\end{aligned}\tag{B.2}$$

Combining (B.2) and (B.1) we deduce that

$$\begin{aligned}
\{A_i(X_i, \boldsymbol{\eta}) R_i(\boldsymbol{\eta})\}_s &= \sum_{r=1}^q a_{isr}(X_i, \boldsymbol{\eta}) \{\bar{Y}_i^r - Q_{ir}(\boldsymbol{\eta})\} \\
&= \{A_i(X_i, \boldsymbol{\eta}^0) R_i(\boldsymbol{\eta}^0)\}_s + \sum_{r=1}^q \left[ \{\bar{Y}_i^r - Q_{ir}(\boldsymbol{\eta}^0)\} a'_{isr}(X_i, \boldsymbol{\eta}^0) - a_{isr}(X_i, \boldsymbol{\eta}^0) u_{ir} \right]^\top \Delta \\
&\quad + \Theta_{is}(\boldsymbol{\eta}) (|\bar{Y}_i^q| + 1) \|\boldsymbol{\eta} - \boldsymbol{\eta}^0\|^2.
\end{aligned}$$

(If  $a_{isr}(x, \boldsymbol{\eta})$  does not actually depend on  $\boldsymbol{\eta}$  then derivatives of  $a_{isr}$  with respect  $\boldsymbol{\eta}$  vanish, and so the term in  $|\bar{Y}_i^r|$  can be dropped. This is the reason it is not necessary, when  $A_i(x, \boldsymbol{\eta})$  does not depend on  $\boldsymbol{\eta}$ , to have  $E(|Y|^c | X = x)$  bounded in  $x$  for all values of  $c$ , although boundedness of  $E(|Y|^p | X = x)$  is needed in order to work with  $q$ th cumulants.) Summing over  $i$ , and equating to zero as entailed by (2.12), we deduce that  $\boldsymbol{\eta} = \hat{\boldsymbol{\eta}}$  satisfies

$$\begin{aligned}
&\frac{1}{k} \sum_{i=1}^k \sum_{r=1}^q \left[ a_{isr}(X_i, \boldsymbol{\eta}^0) u_{ir} - \{\bar{Y}_i^r - Q_{ir}(\boldsymbol{\eta}^0)\} a'_{isr}(X_i, \boldsymbol{\eta}^0) \right]^\top (\boldsymbol{\eta} - \boldsymbol{\eta}^0) \\
&+ \Theta_s(\boldsymbol{\eta}) (|\bar{Y}^r| + 1) \|\boldsymbol{\eta} - \boldsymbol{\eta}^0\|^2 = \frac{1}{k} \sum_{i=1}^k \{A_i(X_i, \boldsymbol{\eta}^0) R_i(\boldsymbol{\eta}^0)\}_s, \tag{B.3}
\end{aligned}$$

or equivalently,

$$(M + L) (\boldsymbol{\eta} - \boldsymbol{\eta}^0) + \Theta(\boldsymbol{\eta}) \|\boldsymbol{\eta} - \boldsymbol{\eta}^0\|^2 = S. \tag{B.4}$$

In (B.3),  $\Theta_s$  is a random variable satisfying, with probability 1,  $|\Theta_s(\boldsymbol{\eta})| \leq p^{-1/2} C_1$  for  $\|\boldsymbol{\eta} - \boldsymbol{\eta}^0\| \leq C_2$  and  $C_2$  sufficiently small, where  $C_1$  depends on  $C_2$ , and in (B.4),  $\Theta(\boldsymbol{\eta})$  is the  $p$ -vector with  $s$ th component  $\Theta_s(\boldsymbol{\eta})$ .

In view of (4.5), the summands in the definition of  $b_{st}$  at (4.3) are uniformly bounded, and all moments of the summands in (4.4) are finite. Moreover,  $\ell_{rj}$  has zero mean and variance

of order  $k^{-1}$ . Therefore Rosenthal's inequality implies that for all  $C_3, C_4 > 1$ ,

$$P(\|L\| > C_3) \leq C_5(C_4) (C_3 k^{1/2})^{-C_4}, \quad (\text{B.5})$$

where  $C_5(C_4) > 0$  depends on  $C_4$  but not on  $C_3$  or  $k$ . (In (B.5), and below, we write the norm  $\|Q\|$  of a  $p \times p$  matrix to denote the supremum of  $\|Qv\|$  over all  $p$ -vectors  $v$  for which  $\|v\| = 1$ .)

Let  $c \in (0, 1)$  denote a lower bound to the least eigenvalue of  $M^T M$ , and put  $C_3 = \frac{1}{3} c$ . If  $\|\boldsymbol{\eta} - \boldsymbol{\eta}^0\| \leq \min(C_1^{-1} C_3, C_2)$  and  $\|L\| \leq C_3$  then

$$\|(M + L)(\boldsymbol{\eta} - \boldsymbol{\eta}^0)\| - \|\Theta(\boldsymbol{\eta})\| \|\boldsymbol{\eta} - \boldsymbol{\eta}^0\|^2 \geq (c - \|L\| - C_3) \|\boldsymbol{\eta} - \boldsymbol{\eta}^0\| \geq C_3 \|\boldsymbol{\eta} - \boldsymbol{\eta}^0\|,$$

and therefore equations (2.12) have a solution whenever

$$\|S\| \leq C_3 \min(C_1^{-1} C_3, C_2).$$

Hence, the probability of a solution is not less than  $\pi_k \equiv 1 - C_6 k^{-C_4/2}$ , for some  $C_6 > 0$ . Moreover, we can deduce from (B.4) that under the same conditions, any solution  $\hat{\boldsymbol{\eta}}$  of (B.1) satisfies

$$\left\| (\hat{\boldsymbol{\eta}} - \boldsymbol{\eta}^0) - (M + L)^{-1} S \right\| \leq (3/2c) C_1 \|\hat{\boldsymbol{\eta}} - \boldsymbol{\eta}^0\|^2, \quad (\text{B.6})$$

and therefore, if  $\|\hat{\boldsymbol{\eta}} - \boldsymbol{\eta}^0\| \leq C_3$ , we can see from (B.4) that

$$\frac{1}{2} \|\hat{\boldsymbol{\eta}} - \boldsymbol{\eta}^0\| \leq \|\hat{\boldsymbol{\eta}} - \boldsymbol{\eta}^0\| \{1 - (3/2c) \|\hat{\boldsymbol{\eta}} - \boldsymbol{\eta}^0\|\} \leq \|(M + L)^{-1} S\| \leq (3/2c) \|S\|. \quad (\text{B.7})$$

Together, (B.6) and (B.7) imply that if  $\|\hat{\boldsymbol{\eta}} - \boldsymbol{\eta}^0\| \leq \min(C_1^{-1} C_3, C_2, C_3)$  then with probability

not less than  $\pi_k$ ,

$$\left\| (\hat{\boldsymbol{\eta}} - \boldsymbol{\eta}^0) - (M + L)^{-1} S \right\| \leq (3/2c) C_1 C_3^{-2} \|S\|^2,$$

from which follows the second-last inequality in the Theorem. The last inequality in the theorem follows via a Taylor expansion, enabled by (B.5).

## B.2 Proof of Corollary 4.1

Define the event  $\mathcal{E} = \{\|S\| \leq \epsilon, \|L\| \leq \epsilon\}$ , and let  $\tilde{\mathcal{E}}$  denote the complement of  $\mathcal{E}$ . Let  $\epsilon > 0$ , let  $C > 0$  be as in the condition  $\max_{1 \leq i \leq k} n_i = O(k^C)$  in the statement of the corollary, and note that

$$\begin{aligned} (\sup |f|)^{-1} E\{|f(\hat{\boldsymbol{\eta}})| I(\tilde{\mathcal{E}})\} &\leq P(\tilde{\mathcal{E}}) \leq P(\|S\| > \epsilon) + P(\|L\| > \epsilon) \\ &\leq \epsilon^{-2} E(\|S\|^2) + \epsilon^{-2(C+1)} E(\|L\|^{2(C+1)}) \\ &= O(K^{-1} + k^{-(C+1)}) = O(K^{-1}). \end{aligned} \quad (\text{B.8})$$

Assume that  $f$  has two bounded derivatives within radius  $C_7$  of  $\boldsymbol{\eta}^0$ , where  $C_7 > 0$ , and let  $D_4$  be as in Theorem 4.1. Let  $c \in (0, 1)$  be a lower bound to the least eigenvalue of  $M^T M$ , and write  $f'$  for the  $p$ -vector of first derivatives of  $f$ . By choosing  $\epsilon = \epsilon(C_7)$  sufficiently small we can show, as in the proof of Theorem 4.1, that when  $\mathcal{E}$  obtains we have  $\|\hat{\boldsymbol{\eta}} - \boldsymbol{\eta}^0\| \leq \frac{1}{2} C_7$ ,  $\|(M + L)^{-1} S\|^T f'(\boldsymbol{\eta}^0) \leq \frac{1}{2} C_7$  and  $D_4 \|S\|^2 \leq \frac{1}{2} C_7$ , and that the least eigenvalues of  $(M + L)^T(M + L)$  and  $(I + LM^{-1})^T(I + LM^{-1})$  exceed  $\frac{1}{2} c$ . Hence, by Taylor expansion of  $f(\hat{\boldsymbol{\eta}})$  about  $\boldsymbol{\eta}^0$ , we deduce from Theorem 4.1 that there exists a constant  $C_8 > 0$  such that, provided  $\mathcal{E}$  obtains,

$$\left| f(\hat{\boldsymbol{\eta}}) - f(\boldsymbol{\eta}^0) - \{(M + L)^{-1} S\}^T f'(\boldsymbol{\eta}^0) \right| \leq C_8 \|S\|^2.$$

Therefore,

$$\begin{aligned} & \left| E\{f(\hat{\boldsymbol{\eta}}) I(\mathcal{E})\} - f(\boldsymbol{\eta}^0) P(\mathcal{E}) - E\left[\{(M+L)^{-1} S\}^T I(\mathcal{E})\right] f'(\boldsymbol{\eta}^0) \right| \\ &= O(E\|S\|^2) = O(K^{-1}). \end{aligned} \quad (\text{B.9})$$

Let  $j_0 \geq 1$  denote an integer. Noting the properties discussed in the previous paragraph, we deduce that

$$\begin{aligned} & E\{(M+L)^{-1} S I(\mathcal{E})\} = M^{-1} E\left\{(I+LM^{-1})^{-1} S I(\mathcal{E})\right\} \\ &= M^{-1} \sum_{j=0}^{j_0} E\left\{(-LM^{-1})^j S I(\mathcal{E})\right\} + O\left[E\{\|L\|^{j_0+1} \|S\| I(\mathcal{E})\}\right] \\ &= M^{-1} \sum_{j=0}^{j_0} \left[ E\{(-LM^{-1})^j S\} - E\{(-LM^{-1})^j S I(\|S\| > \epsilon)\} \right] \\ & \quad + O\left\{ \sum_{j=0}^{j_0} \left[ E\{\|L\|^{2j} I(\|L\| > \epsilon)\} \right]^{1/2} (E\|S\|^2)^{1/2} + E\{\|L\|^{j_0+1} \|S\| I(\mathcal{E})\} \right\}, \end{aligned} \quad (\text{B.10})$$

where, here and below, order-of-magnitude expressions for vectors are interpreted component by component. We can further expand terms in  $L$ , writing  $L = L_1 + L_2$  where  $L_1$  is the  $p \times p$  matrix with  $(s, t)$ th component equal to  $b_{st} - E(b_{st})$ , and  $L_2 = L - L_1$  (see (4.4) for a definition of  $L$ ). Condition (4.5) implies that  $E\|L_1\|^j = O(k^{-j/2})$  and  $E\|L_2\|^j + E\|S\|^j = O(K^{-j/2})$  for all integers  $j$ , and so we can deduce from (B.10) that the same expansion holds if we replace  $L$  by  $L_1$ , replace  $\epsilon$  by  $\frac{1}{2}\epsilon$  in one place, and add a remainder of  $O(K^{-1} + k^{-(j_0+1)} K^{-1/2})$  to

the right-hand side:

$$\begin{aligned}
& E\{(M + L_1)^{-1} S I(\mathcal{E})\} \\
= & M^{-1} \sum_{j=0}^{j_0} \left[ E\{(-L_1 M^{-1})^j S\} - E\{(-L_1 M^{-1})^j S I(\|S\| > \epsilon)\} \right] \\
& + O\left\{ \sum_{j=0}^{j_0} \left[ E\{\|L_1\|^{2j} I(\|L_1\| > \frac{1}{2}\epsilon)\} \right]^{1/2} (E\|S\|^2)^{1/2} \right. \\
& \left. + E\{\|L_1\|^{j_0+1} \|S\| I(\mathcal{E})\} \right\} + O(K^{-1} + k^{-(j_0+1)} K^{-1/2}). \quad (\text{B.11})
\end{aligned}$$

Since  $E(\|S\|^2) = O(K^{-1})$  then

$$\begin{aligned}
E\{\|L_1\|^{j_0+1} \|S\| I(\mathcal{E})\} & \leq \left[ E(\|L_1\|^{2(j_0+1)}) E(\|S\|^2) \right]^{1/2} \\
& = O\left\{ (E\|L_1\|^{2(j_0+1)})^{1/2} K^{-1/2} \right\}.
\end{aligned}$$

Using the argument leading to (B.8) we can show that if  $j_0$  is sufficiently large,

$$E(\|L_1\|^{2(j_0+1)}) = O(k^{-(j_0+1)}) = O(K^{-1}), \quad (\text{B.12})$$

and therefore

$$E\{\|L_1\|^{j_0+1} \|S\| I(\mathcal{E})\} = O(K^{-1}). \quad (\text{B.13})$$

Moreover,  $E(S | \mathcal{F}_X) = 0$ , where  $\mathcal{F}_X$  denotes the sigma-field generated by  $X_1, X_2, \dots$ , and so

$$E\{(-L_1 M^{-1})^j S\} = E\{(-L_1 M^{-1})^j E(S | \mathcal{F}_X)\} = 0. \quad (\text{B.14})$$

Also, for  $j \geq 1$ ,

$$\begin{aligned} E\{(-L_1 M^{-1})^j S I(\|S\| > \epsilon)\} &= E\left[(-L_1 M^{-1})^j E\{S I(\|S\| > \epsilon) \mid \mathcal{F}_X\}\right] \\ &= O\left[E\{\|L_1\|^j E(\|S\|^2 \mid \mathcal{F}_X)\}\right] = O(K^{-1}). \end{aligned} \quad (\text{B.15})$$

Furthermore, using (B.5) we deduce that

$$\begin{aligned} E\{\|L_1\|^{2j} I(\|L_1\| > \tfrac{1}{2} \epsilon)\} &\leq (E\|L_1\|^{4j})^{1/2} P(\|L_1\| > \tfrac{1}{2} \epsilon)^{1/2} \\ &= O(k^{-2j} k^{-C_4/2}), \end{aligned} \quad (\text{B.16})$$

where the last identity holds provided  $C_4 \geq 2(C + 1)$ ; see (B.8). Additionally, by (B.8),

$P(\mathcal{E}) = 1 - P(\tilde{\mathcal{E}}) = 1 - O(K^{-1})$ . Combining this property, (B.9) and (B.11)–(B.15) we deduce that

$$|E\{f(\hat{\boldsymbol{\eta}}) I(\mathcal{E})\} - f(\boldsymbol{\eta}^0)| = O(K^{-1}). \quad (\text{B.17})$$

The corollary follows from (B.8) and (B.17).

### B.3 Proof of (4.8)

Assume that  $\min_{1 \leq i \leq k} n_i(k) \rightarrow \infty$  and that (4.5) holds. Then, writing  $U_{ir}$  for  $\bar{Y}_i^r - Q_{ir}$ , and noting that  $E(S_r | \mathcal{X}) = 0$ , we have:

$$\begin{aligned}
\text{cov}(S_r, S_s) &= E\{\text{cov}(S_r, S_s | \mathcal{X})\} + \text{cov}\{E(S_r | \mathcal{X}) E(S_s | \mathcal{X})\} \\
&= E\{\text{cov}(S_r, S_s | \mathcal{X})\} = \frac{1}{k^2} \sum_{i=1}^k E\{\text{cov}(U_{ir}, U_{is} | X_i)\} \\
&= \frac{1}{k^2} \sum_{i=1}^k E\left\{\text{cov}\left(\left[\lambda_1(X_i, \boldsymbol{\eta}) + \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\}\right]^r, \right. \right. \\
&\quad \left. \left. \left[\lambda_1(X_i, \boldsymbol{\eta}) + \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\}\right]^s \mid X_i\right)\right\} \\
&= \frac{1}{k^2} \sum_{i=1}^k E\left( E\left[ \lambda_1(X_i, \boldsymbol{\eta})^{r+s} + (r+s) \lambda_1(X_i, \boldsymbol{\eta})^{r+s-1} \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\} \right. \right. \\
&\quad \left. \left. + \frac{1}{2} (r+s)(r+s-1) \lambda_1(X_i, \boldsymbol{\eta})^{r+s-2} \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\}^2 \mid X_i \right] \right. \\
&\quad \left. - \left[ E\left\{ \lambda_1(X_i, \boldsymbol{\eta})^r + r \lambda_1(X_i, \boldsymbol{\eta})^{r-1} \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\} \right. \right. \right. \\
&\quad \left. \left. \left. + \frac{1}{2} r(r-1) \lambda_1(X_i, \boldsymbol{\eta})^{r-2} \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\}^2 \mid X_i \right\} \right] \right. \\
&\quad \left. \times \left[ E\left\{ \lambda_1(X_i, \boldsymbol{\eta})^s + s \lambda_1(X_i, \boldsymbol{\eta})^{s-1} \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\} \right. \right. \right. \\
&\quad \left. \left. \left. + \frac{1}{2} s(s-1) \lambda_1(X_i, \boldsymbol{\eta})^{s-2} \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\}^2 \mid X_i \right\} \right] \right) + o(K^{-1}) \\
&= \frac{1}{k^2} \sum_{i=1}^k E\left( \lambda_1(X_i, \boldsymbol{\eta})^{r+s} + \frac{1}{2} (r+s)(r+s-1) \lambda_1(X_i, \boldsymbol{\eta})^{r+s-2} \right. \\
&\quad \left. \times E[\{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\}^2 \mid X_i] \right. \\
&\quad \left. - \left( \lambda_1(X_i, \boldsymbol{\eta})^r + \frac{1}{2} r(r-1) \lambda_1(X_i, \boldsymbol{\eta})^{r-2} \right. \right. \\
&\quad \left. \left. \times E[\{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\}^2 \mid X_i] \right)^2 \right) \\
&\quad \times \left( \lambda_1(X_i, \boldsymbol{\eta})^s + \frac{1}{2} s(s-1) \lambda_1(X_i, \boldsymbol{\eta})^{s-2} \right. \\
&\quad \left. \times E[\{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\}^2 \mid X_i] \right)^2 + o(K^{-1}) \\
&= \frac{1}{2k^2} \{(r+s)(r+s-1) - r(r-1) - s(s-1)\} \\
&\quad \times \sum_{i=1}^k n_i^{-1} E\left\{ \lambda_1(X_i, \boldsymbol{\eta})^{r+s-2} \text{var}(Y_i | X_i) \right\} + o(K^{-1}) \\
&= rs K^{-1} E\left\{ \lambda_1(X, \boldsymbol{\eta})^{r+s-2} \text{var}(Y | X) \right\} + o(K^{-1}),
\end{aligned}$$

which establishes (4.8).

## B.4 Proof of Theorem 4.2

The version, for the case of centred moments, of the argument leading to (B.3) is identical in essential details to that given before, and shows that  $\hat{\boldsymbol{\eta}}$  is the solution in  $\boldsymbol{\eta}$  of the equation:

$$\begin{aligned} \frac{1}{k} \sum_{i=1}^k \sum_{r=1}^q \left\{ a'_{isr}(X_i, \boldsymbol{\eta}^0) \check{R}_{ir}(\boldsymbol{\eta}^0) + a_{isr}(X_i, \boldsymbol{\eta}^0) \check{R}'_{ir}(\boldsymbol{\eta}^0) \right\}^T (\boldsymbol{\eta} - \boldsymbol{\eta}^0) \\ + \Theta_s(\boldsymbol{\eta}) (|\bar{Y}^r| + 1) \|\boldsymbol{\eta} - \boldsymbol{\eta}^0\|^2 = -\frac{1}{k} \sum_{i=1}^k \{A_i(X_i, \boldsymbol{\eta}^0) \check{R}_i(\boldsymbol{\eta}^0)\}_s, \end{aligned}$$

where  $\check{R}'_{ir} = (\partial/\partial \boldsymbol{\eta}) \check{R}_{ir}$  is a  $p$ -vector. Equivalently, in place of (B.4),

$$(\check{M} + \check{L})(\boldsymbol{\eta} - \boldsymbol{\eta}^0) + \Theta(\boldsymbol{\eta}) \|\boldsymbol{\eta} - \boldsymbol{\eta}^0\|^2 = \check{S}.$$

The remainder of the proof is virtually equivalent to that of Theorem 4.1.

## B.5 Proof of nonsingularity of $A(X_i, \boldsymbol{\eta})$ when defined by (2.15) and (4.6) holds

For brevity we treat the case where  $p = q = 2$  and  $n_i \equiv n$ , although other cases are similar. Writing  $\mu_{ir} = E[\{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\}^r | X_i]$ , the first of the following results can be deduced from (2.16):

$$\boldsymbol{\Sigma}_i = \begin{pmatrix} \mu_{i2} & \mu_{i3} \\ \mu_{i3} & \mu_{i4} - \mu_{i2}^2 \end{pmatrix}, \boldsymbol{\Sigma}_i^{-1} = |\boldsymbol{\Sigma}_i|^{-1} \begin{pmatrix} \mu_{i4} - \mu_{i2}^2 & -\mu_{i3} \\ -\mu_{i3} & \mu_{i2} \end{pmatrix},$$

$$\lambda'_r(X_i, \boldsymbol{\eta}) = \begin{pmatrix} \lambda'_{r1}(X_i, \boldsymbol{\eta}) \\ \lambda'_{r2}(X_i, \boldsymbol{\eta}) \end{pmatrix}, \check{\mathbf{D}}_i = \begin{pmatrix} \lambda'_{11}(X_i, \boldsymbol{\eta}) & n^{-1} \lambda'_{21}(X_i, \boldsymbol{\eta}) \\ \lambda'_{12}(X_i, \boldsymbol{\eta}) & n^{-1} \lambda'_{22}(X_i, \boldsymbol{\eta}) \end{pmatrix},$$

where  $\lambda'_r(X_i, \boldsymbol{\eta})$  is the 2-vector of derivatives of  $\lambda_r(X_i, \boldsymbol{\eta})$  with respect to  $\boldsymbol{\eta}$ ,  $\lambda'_{rj} = \partial \lambda_r / \partial \eta_j$  and we have used the fact that  $W_{ir} = n^{1-r}$ . Similarly,  $\mu_{i2} = n^{-1} \lambda_2(X_i, \boldsymbol{\eta})$ ,  $\mu_{i3} = n^{-2} \lambda_3(X_i, \boldsymbol{\eta})$ ,

$$\begin{aligned} \mu_{i4} - \mu_{i2}^2 &= n^{-3} \lambda_4(X_i, \boldsymbol{\eta}) + 3 \{n^{-1} \lambda_2(X_i, \boldsymbol{\eta})\}^2 - \{n^{-1} \lambda_2(X_i, \boldsymbol{\eta})\}^2 \\ &= n^{-3} \lambda_4(X_i, \boldsymbol{\eta}) + 2 \{n^{-1} \lambda_2(X_i, \boldsymbol{\eta})\}^2, \end{aligned}$$

$$|\boldsymbol{\Sigma}_i| = \mu_{i2} \mu_{i4} - \mu_{i2}^3 - \mu_{i3}^2 = 2 \mu_{i2}^3 + O_p(n^{-4}) = 2 n^{-3} \lambda_2(X_i, \boldsymbol{\eta})^3 + O_p(n^{-4}).$$

Therefore, provided that  $\lambda_2(x, \boldsymbol{\eta})$  is bounded away from zero uniformly in  $x$  in the support of the distribution of  $X$  and in a neighbourhood of  $\boldsymbol{\eta}^0$ , we have:

$$\boldsymbol{\Sigma}_i^{-1} = \frac{1 + O_p(n^{-1})}{2 n^{-3} \lambda_2^3} \begin{pmatrix} n^{-3} \lambda_4 + 2 n^{-2} \lambda_2^2 & -n^{-2} \lambda_3 \\ -n^{-2} \lambda_3 & n^{-1} \lambda_2 \end{pmatrix}.$$

Here and below we write simply  $\lambda$  and  $\lambda'_{rj}$  for  $\lambda(X_i, \boldsymbol{\eta})$  and  $\lambda'_{rj}(X_i, \boldsymbol{\eta})$ , respectively. Hence, defining  $\lambda_r(X_i, \boldsymbol{\eta}) = (\lambda_{r1}(X_i, \boldsymbol{\eta}), \lambda_{r2}(X_i, \boldsymbol{\eta}))^T$ , a 2-vector, we deduce that

$$n^{-1} \check{\mathbf{D}}_i \boldsymbol{\Sigma}_i^{-1} = \frac{1 + O_p(n^{-1})}{2 \lambda_2^3} \begin{pmatrix} 2 \lambda_2^2 \lambda'_{11} + n^{-1} (\lambda_4 \lambda'_{11} - \lambda_3 \lambda'_{21}) & \lambda_2 \lambda'_{21} - \lambda_3 \lambda'_{11} \\ 2 \lambda_2^2 \lambda'_{12} + n^{-1} (\lambda_4 \lambda'_{12} - \lambda_3 \lambda'_{22}) & \lambda_2 \lambda'_{22} - \lambda_3 \lambda'_{12} \end{pmatrix} = \begin{pmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{pmatrix},$$

say, where  $\lambda_r$  and  $\lambda'_{rj}$  denote  $\lambda_r(X_i, \boldsymbol{\eta})$  and  $\lambda'_{rj}(X_i, \boldsymbol{\eta})$ , respectively, and we have suppressed the dependence of these quantities and the coefficients  $c_{st}$  on  $i$ .

Since  $\check{R}_{ir} = \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\}^r - \check{Q}_{ir}$  then

$$\begin{aligned} -\check{R}'_{ir} &= -\partial \frac{\partial}{\partial \boldsymbol{\eta}} \check{R}_{ir} = r \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\}^{r-1} \lambda'_1 + \partial \frac{\partial}{\partial \boldsymbol{\eta}} \check{Q}_{ir} \\ &= r \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\}^{r-1} \begin{pmatrix} \lambda'_{11} \\ \lambda'_{12} \end{pmatrix} + \begin{cases} 0 & \text{if } r = 1 \\ n^{-1} \begin{pmatrix} \lambda_{21} \\ \lambda_{22} \end{pmatrix} & \text{if } r = 2, \end{cases} \end{aligned}$$

because  $\check{Q}_{i1} \equiv 0$ . Therefore, writing  $\Delta = (\Delta_1, \Delta_2)^T = \boldsymbol{\eta} - \boldsymbol{\eta}^0$ , we have:

$$\begin{aligned} \check{R}_{ir}(\boldsymbol{\eta}) &= \check{R}_{ir}(\boldsymbol{\eta}^0) + \check{R}'_{ir}(\boldsymbol{\eta}^0)^T (\boldsymbol{\eta} - \boldsymbol{\eta}^0) + O_p(\|\Delta\|^2) \\ &= \check{R}_{ir}(\boldsymbol{\eta}^0) - r \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\}^{r-1} (\lambda'_{11} \Delta_1 + \lambda'_{12} \Delta_2) \\ &\quad + O_p(\|\Delta\|^2) - \begin{cases} 0 & \text{if } r = 1 \\ n^{-1} (\lambda'_{21} \Delta_1 + \lambda'_{22} \Delta_2) & \text{if } r = 2. \end{cases} \end{aligned}$$

Hence,

$$\begin{aligned} &n^{-1} \sum_{i=1}^k \check{D}_i \boldsymbol{\Sigma}_i^{-1} \check{R}_i \\ &= \sum_{i=1}^k \begin{pmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{pmatrix} \left\{ \check{R}_i(\boldsymbol{\eta}^0) - \begin{pmatrix} \lambda'_{11} \Delta_1 + \lambda'_{12} \Delta_2 \\ 2 \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\} (\lambda'_{11} \Delta_1 + \lambda'_{12} \Delta_2) + n^{-1} (\lambda'_{21} \Delta_1 + \lambda'_{22} \Delta_2) \end{pmatrix} \right\} \\ &+ O_p(k \|\Delta\|^2) \\ &= \sum_{i=1}^k \left\{ \begin{pmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{pmatrix} \check{R}_i(\boldsymbol{\eta}^0) - \begin{pmatrix} c_{11} (\lambda'_{11} \Delta_1 + \lambda'_{12} \Delta_2) + c_{12} \delta \\ c_{21} (\lambda'_{11} \Delta_1 + \lambda'_{12} \Delta_2) + c_{22} \delta \end{pmatrix} \right\} + (\text{negligible}), \end{aligned}$$

where

$$\delta = 2 \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\} (\lambda'_{11} \Delta_1 + \lambda'_{12} \Delta_2) + n^{-1} (\lambda'_{21} \Delta_1 + \lambda'_{22} \Delta_2).$$

Now,

$$\begin{aligned} & \frac{1}{k} \sum_{i=1}^k (c_{j1} \check{R}_{i1} + c_{j2} \check{R}_{i2}) \\ &= \frac{1}{k} \sum_{i=1}^k \left( c_{j1}(i) \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\} + c_{j2}(i) [\{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\}^2 - n^{-1} \lambda_2(X_i, \boldsymbol{\eta})] \right). \end{aligned}$$

The two terms here give contributions of sizes  $(kn)^{-1/2}$  and  $k^{-1/2}n^{-1}$ , respectively, so the second is negligible. Now,

$$a_1 = a_1(i) \equiv c_{11} \lambda'_{11} = \lambda_2^{-1} \lambda'_{11}{}^2 + O(n^{-1}),$$

$$a_2 = a_2(i) \equiv c_{11} \lambda'_{12} = \lambda_2^{-1} \lambda'_{11} \lambda'_{12} + O(n^{-1}),$$

$$c_{21} \lambda'_{11} = \lambda_2^{-1} \lambda'_{11} \lambda'_{12} + O(n^{-1}), \quad a_3 = a_3(i) \equiv c_{21} \lambda'_{12} = \lambda_2^{-1} \lambda'_{12}{}^2 + O(n^{-1}),$$

and so

$$a_2 = a_2(i) = c_{11} \lambda'_{12} = c_{21} \lambda'_{11} + O(n^{-1}).$$

Therefore,

$$n^{-1} \sum_{i=1}^k \check{D}_i \boldsymbol{\Sigma}_i^{-1} \check{R}_i = \sum_{i=1}^k \left\{ \begin{pmatrix} c_{11}(i) \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\} \\ c_{21}(i) \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\} \end{pmatrix} - \begin{pmatrix} a_1 & a_2 \\ a_2 & a_3 \end{pmatrix} \begin{pmatrix} \Delta_1 \\ \Delta_2 \end{pmatrix} \right\} + (\text{negligible}).$$

The condition on  $M^{0T}M^0$  being nonsingular reduces to asymptotic nonsingularity of the matrix

$$\frac{1}{k} \sum_{i=1}^k \begin{pmatrix} a_1 & a_2 \\ a_2 & a_3 \end{pmatrix}$$

Therefore: the first of the following holds:

$$\begin{aligned} \sum_{i=1}^k (c_{11} \check{R}_{i1} + n c_{12} \check{R}_{i2}) &= \sum_{i=1}^k \left\{ c_{11} (\lambda_{11} \Delta_1 + \lambda_{12} \Delta_2) + c_{12} (\lambda_{21} \Delta_1 + \lambda_{22} \Delta_2) \right\} + (\text{negligible}), \\ \sum_{i=1}^k (c_{21} \check{R}_{i1} + n c_{22} \check{R}_{i2}) &= \sum_{i=1}^k \left\{ c_{21} (\lambda_{11} \Delta_1 + \lambda_{12} \Delta_2) + c_{22} (\lambda_{21} \Delta_1 + \lambda_{22} \Delta_2) \right\} + (\text{negligible}), \end{aligned}$$

where

$$\check{R}_{i1} = \bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta}), \quad \check{R}_{i2} = \{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\}^2 - E[\{\bar{Y}_i - \lambda_1(X_i, \boldsymbol{\eta})\}^2 | X_i].$$

We can solve this for  $\Delta_1$  and  $\Delta_2$ , obtaining the result that  $\Delta_1$  and  $\Delta_2$  both equal linear combinations in  $k^{-1} \sum_i \check{R}_{i1}$  and  $n k^{-1} \sum_i \check{R}_{i2}$ . The second of these is only  $O(k^{-1})$ .