

SUPPLEMENTARY MATERIAL

6. NOTATION

Let $N(x; \mu, \Sigma)$ be the normal density at x with mean μ and variance Σ , and $\tilde{f}_n(s | \theta) = N\{s; s(\theta), A(\theta)/a_n^2\}$, the asymptotic distribution of the summary statistic. We define $a_{n,\varepsilon} = a_n$ if $\lim_{n \rightarrow \infty} a_n \varepsilon_n < \infty$ and $a_{n,\varepsilon} = \varepsilon_n^{-1}$ otherwise, and $c_\varepsilon = \lim_{n \rightarrow \infty} a_n \varepsilon_n$, both of which summarize how ε_n decreases relative to the converging rate, a_n , of S_n in Condition 2 below. Define the standardized random variables $W_n(S_n) = a_n A(\theta)^{-1/2} \{S_n - s(\theta)\}$ and $W_{\text{obs}} = a_n A(\theta)^{-1/2} \{s_{\text{obs}} - s(\theta_0)\}$ and $\beta_0 = I(\theta_0)^{-1} Ds(\theta_0)^T A(\theta_0)^{-1}$ according to Condition 2 below.

Let $f_{W_n}(w | \theta)$ and $\tilde{f}_{W_n}(w | \theta)$ be the density for $W_n(S_n)$ when $S_n \sim f_n(\cdot | \theta)$ and $\tilde{f}_n(\cdot | \theta)$ respectively. Let $B_\delta = \{\theta | \|\theta - \theta_0\| \leq \delta\}$ for $\delta > 0$. Define the initial density truncated in B_δ , i.e. $r_n(\theta) \mathbb{I}_{\theta \in B_\delta} / \int_{B_\delta} r_n(\theta) d\theta$, by $r_\delta(\theta)$. Let $t(\theta) = a_{n,\varepsilon}(\theta - \theta_0)$ and $v(s) = \varepsilon_n^{-1}(s - s_{\text{obs}})$. For any $A \in \mathcal{B}^p$ where \mathcal{B}^p is the Borel sigma-field on \mathbb{R}^p , let $t(A)$ be the set $\{\phi : \phi = t(\theta) \text{ for some } \theta \in A\}$. For a non-negative function $h(x)$, integrable in \mathbb{R}^l , denote the normalized function $h(x) / \int_{\mathbb{R}^l} h(x) dx$ by $h(x)^{(\text{norm})}$. For a function $h(x)$, denote its gradient by $D_x h(x)$, and for simplicity, omit θ from D_θ . For a sequence x_n , we use the notation $x_n = \Theta(a_n)$ to mean that there exist some constants m and M such that $0 < m < |x_n/a_n| < M < \infty$.

7. ADDITIONAL CONDITIONS

Condition 7. The kernel satisfies

- (i) $\int v K_\varepsilon(v) dv = 0$;
- (ii) $\prod_{k=1}^l v_{i_k} K_\varepsilon(v) dv < \infty$ for any coordinates $(v_{i_1}, \dots, v_{i_l})$ of v and $l \leq p + 6$;
- (iii) $K_\varepsilon(v) \propto K_\varepsilon(\|v\|_\Lambda^2)$ where $\|v\|_\Lambda^2 = v^T \Lambda v$ and Λ is a positive-definite matrix, and $K(v)$ is a decreasing function of $\|v\|_\Lambda$; (iv) $K_\varepsilon(v) = O(\exp\{-c_1 \|v\|^{\alpha_1}\})$ for some $\alpha_1 > 0$ and $c_1 > 0$ as $\|v\| \rightarrow \infty$.

Condition 8. There exists α_n satisfying $\alpha_n/a_n^{2/5} \rightarrow \infty$ and a density $r_{\max}(w)$ satisfying Condition 7(ii)–(iii) where $K_\varepsilon(v)$ is replaced with $r_{\max}(w)$, such that $\sup_{\theta \in B_\delta} \alpha_n |f_{W_n}(w | \theta) - \tilde{f}_{W_n}(w | \theta)| \leq c_3 r_{\max}(w)$ for some positive constant c_3 .

Condition 9. The following statements hold:

- (i) $r_{\max}(w)$ satisfies Condition 7(iv); and
- (ii) $\sup_{\theta \in B_\delta^c} \tilde{f}_{W_n}(w | \theta) = O(e^{-c_2 \|w\|^{\alpha_2}})$ as $\|w\| \rightarrow \infty$ for some positive constants c_2 and α_2 , and $A(\theta)$ is bounded in \mathcal{P} .

Condition 10. The first two moments, $\int_{\mathbb{R}^d} s \tilde{f}_n(s | \theta) ds$ and $\int_{\mathbb{R}^d} s^T s \tilde{f}_n(s | \theta) ds$, exist.

8. PROOF FOR THEOREM 2

Let

$$\tilde{Q}(\theta \in A | s) = \int_A r_\delta(\theta) \tilde{f}_n(s | \theta) d\theta / \int_{\mathbb{R}^p} r_\delta(\theta) \tilde{f}_n(s | \theta) d\theta.$$

Lemma 2. Assume Condition 3–8. If $\varepsilon_n = O(a_n^{-1})$, for any fixed $\nu \in \mathbb{R}^d$ and small enough δ ,

$$\sup_{A \in \mathcal{B}^p} |\tilde{Q}\{a_n(\theta - \theta_0) \in A | s_{\text{obs}} + \varepsilon_n \nu\} - \int_A N[t; \beta_0 \{A(\theta_0)^{1/2} W_{\text{obs}} + c_\varepsilon \nu\}, I(\theta_0)^{-1}] dt| \rightarrow 0,$$

in probability as $n \rightarrow \infty$, where $\beta_0 = I(\theta_0)^{-1} Ds(\theta_0)^T A(\theta_0)^{-1}$.

Proof of Lemma 2: This result generalizes Lemma A1 in [1]. With Lemma A1 from [1], it is sufficient to show that

$$\sup_{A \in \mathcal{B}^p} |\tilde{Q}\{t(\theta) \in A | s_{\text{obs}} + \varepsilon_n \nu\} - \tilde{\Pi}\{t(\theta) \in A | s_{\text{obs}} + \varepsilon_n \nu\}|$$

is $o_P(1)$ where $\tilde{\Pi}$ denotes the posterior distribution with prior $\pi_\delta(\theta)$ and likelihood $\tilde{f}_n(s | \theta)$. With the transformation $t = t(\theta)$ and $v = v(s)$, the left hand side of the above equation can be written as

$$\sup_{A \in \mathcal{B}^p} \left| \frac{\int_A r_\delta(\theta_0 + a_n^{-1}t) \tilde{f}_n(s_{\text{obs}} + \varepsilon_n \nu | \theta_0 + a_n^{-1}t) dt}{\int_{\mathbb{R}^p} r_\delta(\theta_0 + a_n^{-1}t) \tilde{f}_n(s_{\text{obs}} + \varepsilon_n \nu | \theta_0 + a_n^{-1}t) dt} - (8.1) \right. \\ \left. \frac{\int_A \pi_\delta(\theta_0 + a_n^{-1}t) \tilde{f}_n(s_{\text{obs}} + \varepsilon_n \nu | \theta_0 + a_n^{-1}t) dt}{\int_{\mathbb{R}^p} \pi_\delta(\theta_0 + a_n^{-1}t) \tilde{f}_n(s_{\text{obs}} + \varepsilon_n \nu | \theta_0 + a_n^{-1}t) dt} \right|.$$

For a function $\tau : \mathbb{R}^p \rightarrow \mathbb{R}$, define the following auxiliary functions,

$$\phi_1\{\tau(\theta); n\} = \frac{\int_{t(B_\delta)} |\tau(\theta_0 + a_n^{-1}t) - \tau(\theta)| \tilde{f}_n(s_{\text{obs}} + \varepsilon_n \nu | \theta_0 + a_n^{-1}t) dt}{\int_{t(B_\delta)} \tau(\theta_0 + a_n^{-1}t) \tilde{f}_n(s_{\text{obs}} + \varepsilon_n \nu | \theta_0 + a_n^{-1}t) dt}, \\ \phi_2\{\tau(\theta); n\} = \frac{\tau(\theta) \int_{t(B_\delta)} \tilde{f}_n(s_{\text{obs}} + \varepsilon_n \nu | \theta_0 + a_n^{-1}t) dt}{\int_{t(B_\delta)} \tau(\theta_0 + a_n^{-1}t) \tilde{f}_n(s_{\text{obs}} + \varepsilon_n \nu | \theta_0 + a_n^{-1}t) dt}.$$

Then by adding and subtracting $\phi_2\{\tau_n^{-p} r_\delta(\theta); n\} \phi_2\{\pi(\theta); n\}$ in the absolute sign of (8.1), (8.1) can be bounded by

$$\phi_1\{\tau_n^{-p} r_\delta(\theta); n\} + \phi_1\{\pi(\theta); n\} \phi_2\{\tau_n^{-p} r_\delta(\theta); n\} \\ + \phi_1\{\tau_n^{-p} r_\delta(\theta); n\} \phi_2\{\pi(\theta); n\} + \phi_1\{\pi(\theta); n\}.$$

Consider a class of function $\tau(\theta)$ satisfying the following conditions:

- (1) There exists a series $\{k_n\}$, such that $\sup_{\theta \in \mathcal{P}_0} \|k_n^{-1} D\tau(\theta)\| < \infty$ and $k_n = o(a_n)$;
- (2) $\tau(\theta_0) > 0$ and $\tau(\theta) \in C^1(B_\delta)$.

By Conditions 3–6, $\tau_n^{-p} r_\delta(\theta)$ and $\pi_\delta(\theta)$ belong to the above class. Then if $\phi_1\{\tau(\theta); n\}$ is $o_P(1)$ and $\phi_2\{\tau(\theta); n\}$ is $O_P(1)$, (8.1) is $o_P(1)$ and the lemma holds.

First, from the properties of $\tau(\theta)$ mentioned above, there exists an open set $\omega \subset B_\delta$ such that $\inf_{\theta \in \omega} \tau(\theta) > c_1$, for a

constant $c_1 > 0$. Then for $\phi_2\{\tau(\theta); n\}$, it is bounded by

$$\frac{\tau(\theta)}{c_1 \int_{t(\omega)} \tilde{f}_n(s_{\text{obs}} + \varepsilon_n \nu \mid \theta_0 + a_n^{-1} t)^{(norm)} dt},$$

where $h(x)^{(norm)}$ represents the normalized version of $h(x)$. From equation (7) in the supplementary material of [2], $\tilde{f}_n(s_{\text{obs}} + \varepsilon_n \nu \mid \theta_0 + a_n^{-1} t)$ can be written in the following form,

$$\frac{a_n^d \tilde{f}_n(s_{\text{obs}} + \varepsilon_n \nu \mid \theta_0 + a_n^{-1} t)}{\|B_n(t)\|^{1/2}} N[C_n(t)\{A_n(t)t - b_n \nu - c_2; \theta_0, I_d\}], \quad (8.2)$$

where $A_n(t)$ is a series of $d \times p$ matrix functions, $\{B_n(t)\}$ and $\{C_n(t)\}$ are a series of $d \times d$ matrix functions, b_n converges to a non-negative constant and c_2 is a constant, and the minimum of absolute eigenvalues of $A_n(t)$ and eigenvalues of $B_n(t)$ and $C_n(t)$ are all bounded and away from 0. Then for fixed ν , by continuous mapping, (8.2) is away from zero with probability one. Therefore $\phi_2\{\tau(\theta); n\} = O_P(1)$.

Second, by Taylor expansion, $\tau(\theta_0 + a_n^{-1} t) = \tau(\theta_0) + a_n^{-1} D\tau(\theta_0 + e_t t)t$, where $\|e_t\| \leq a_n^{-1}$. Then $\phi_1\{\tau(\theta); n\}$ is equal to

$$\begin{aligned} & \frac{k_n \phi_2\{\tau(\theta); n\}}{a_n \tau(\theta)} \frac{\int_{t(B_\delta)} |k_n^{-1} D\tau(\theta_0 + e_t t)t| \tilde{f}_n(s_{\text{obs}} + \varepsilon_n \nu \mid \theta_0 + a_n^{-1} t) dt}{\int_{t(B_\delta)} \tilde{f}_n(s_{\text{obs}} + \varepsilon_n \nu \mid \theta_0 + a_n^{-1} t) dt} \\ & \leq \\ & \left(\frac{k_n \phi_2\{\tau(\theta); n\}}{a_n \tau(\theta)} \sup_{\theta \in B_\delta} \|k_n^{-1} D\tau(\theta)\| \right) \times \\ & \left(\frac{\int_{t(B_\delta)} \|t\| a_n^d \tilde{f}_n(s_{\text{obs}} + \varepsilon_n \nu \mid \theta_0 + a_n^{-1} t) dt}{\int_{t(B_\delta)} a_n^d \tilde{f}_n(s_{\text{obs}} + \varepsilon_n \nu \mid \theta_0 + a_n^{-1} t) dt} \right), \quad (8.3) \end{aligned}$$

where the inequality holds by the triangle inequality. By the expression (8.2) and Lemma 7 in the supplementary material of [2], the right hand side of (8.3) is $O_P(1)$. Then together with $\phi_2\{\tau(\theta); n\} = \Theta_P(1)$, $\phi_1\{\tau(\theta); n\} = o_P(1)$. Therefore the Lemma holds. \square

Define the joint density of (θ, s) in Algorithm 1 and its approximation, where the s-likelihood is replaced by its Gaussian limit and $r_n(\theta)$ by its truncation, by $q_\varepsilon(\theta, s)$ and $\tilde{q}_\varepsilon(\theta, s)$. Then

$$\begin{aligned} q_\varepsilon(\theta, s) &= \frac{r_n(\theta) f_n(s \mid \theta) K_{\varepsilon_n}(s - s_{\text{obs}})}{\int_{\mathbb{R}^p \times \mathbb{R}^d} r_n(\theta) f_n(s \mid \theta) K_{\varepsilon_n}(s - s_{\text{obs}}) d\theta ds}, \\ \tilde{q}_\varepsilon(\theta, s) &= \frac{r_\delta(\theta) \tilde{f}_n(s \mid \theta) K_{\varepsilon_n}(s - s_{\text{obs}})}{\int_{\mathbb{R}^p \times \mathbb{R}^d} r_\delta(\theta) \tilde{f}_n(s \mid \theta) K_{\varepsilon_n}(s - s_{\text{obs}}) d\theta ds}. \end{aligned}$$

Let $\tilde{Q}_\varepsilon(\theta \in A \mid s_{\text{obs}})$ be the approximate confidence distribution function equal to $\int_A \int_{\mathbb{R}^d} \tilde{q}_\varepsilon(\theta, s) ds d\theta$. With the transformation $t = t(\theta)$ and $v = v(s)$, let

$$\tilde{q}_{\varepsilon, tv}(t, v) = \tau_n^{-p} r_\delta(\theta_0 + a_{n, \varepsilon}^{-1} t) \tilde{f}_n(s_{\text{obs}} + \varepsilon_n \nu \mid \theta_0 + a_{n, \varepsilon}^{-1} t) K_\varepsilon(\nu)$$

be the transformed and unnormalized $\tilde{q}_\varepsilon(\theta, s)$, and

$$\tilde{q}_{A, tv}(h) = \int_A \int_{\mathbb{R}^d} h(t, v) \tilde{q}_{\varepsilon, tv}(t, v) dv dt$$

for any function $h(\cdot, \cdot)$ in $\mathbb{R}^p \times \mathbb{R}^d$. Denote the factor of $\tilde{q}_{\varepsilon, tv}(t, v)$, $\tau_n^{-p} r_\delta(\theta_0 + a_{n, \varepsilon}^{-1} t)$, by $\gamma_n(t)$. Let $\gamma = \lim_{n \rightarrow \infty} \tau_n^{-p} r_\delta(\theta)$ and $\gamma(t) = \lim_{n \rightarrow \infty} \tau_n^{-p} r_\delta(\theta_0 + \tau_n^{-1} t)$, the limits of $\gamma_n(t)$ when $a_{n, \varepsilon} = a_n$ and $a_{n, \varepsilon} = \tau_n$ respectively. By Condition 4 and 5, $\gamma(t)$ exists and γ is non-zero with positive probability.

Next several functions of t and v defined in [1, proofs for Section 3.1] and relate to the limit of $\tilde{q}_{\varepsilon, tv}(t, v)$ are used, including $g(v; A, B, c)$, $g_n(t, v)$, $G_n(v)$ and $g'_n(t, v)$.

Furthermore, several functions defined by integration as following are used: for any $A \in \mathfrak{B}^p$, let

$$g_{A, r}(h) = \int_{\mathbb{R}^d} \int_{t(A)} h(t, v) \gamma_n(t) g_n(t, v) dt dv,$$

$$G_{n, r}(v) = \int_{t(B_\delta)} \gamma_n(t) g_n(t, v) dt,$$

$$q_A(h) = \int_A \int_{\mathbb{R}^d} h(\theta, s) r_n(\theta) f_n(s \mid \theta) K_\varepsilon(s - s_{\text{obs}}) \varepsilon_n^{-d} ds d\theta,$$

$$\tilde{q}_A(h) = \int_A \int_{\mathbb{R}^d} h(\theta, s) r_\delta(\theta) \tilde{f}_n(s \mid \theta) K_\varepsilon(s - s_{\text{obs}}) \varepsilon_n^{-d} ds d\theta,$$

which generalize those defined in [1, proofs for Section 3.1] for the case $r_n(\theta) = \pi(\theta)$.

Lemma 3. Assume Condition 3–7. If $\varepsilon_n = o(a_n^{-1/2})$, then

- (i) $\int_{\mathbb{R}^d} \int_{t(B_\delta)} |\tilde{q}_{\varepsilon, tv}(t, \nu) - \gamma_n(t) g_n(t, \nu)| dt d\nu = o_p(1)$;
- (ii) $g_{B_\delta, r}(1) = \Theta_P(1)$;
- (iii) $\tilde{q}_{B_\delta, tv}(t^{k_1} v^{k_2}) / \tilde{q}_{B_\delta, tv}(1) = g_{B_\delta, r}(t^{k_1} v^{k_2}) / g_{B_\delta, r}(1) + O_P(a_{n, \varepsilon}^{-1}) + O_P(a_n^2 \varepsilon_n^4)$ for pairs $(k_1, k_2) = (0, 0), (1, 0), (1, 1), (0, 1)$, and $(0, 2)$;
- (iv) $\tilde{q}_{B_\delta}(1) =$

$$\tau_n^p a_{n, \varepsilon}^{d-p} \left\{ \int_{t(B_\delta)} \int_{\mathbb{R}^d} \gamma_n(t) g_n(t, \nu) d\tau d\nu + O_P(a_{n, \varepsilon}^{-1}) + O_P(a_n^2 \varepsilon_n^4) \right\}.$$

Proof of Lemma 3: These results generalize parts of Lemma A2 in [1] (corresponding to items (i) and (ii) above) and Lemma 5 in [2] (corresponding to items (iii) and (iv) above).

To prove part (i), note that in Lemma A2 of [1] $\gamma_n(t) = \pi(\theta_0 + a_{n, \varepsilon}^{-1} t)$, and (i) holds by expanding $\tilde{q}_{\varepsilon, tv}(t, v)$ according to the proof of Lemma 5 of [2]. For $\gamma_n(t) = \tau_n^{-p} r_\delta(\theta_0 + a_{n, \varepsilon}^{-1} t)$ this can be similarly proved by changing the terms involving $\pi(\theta)$ in equations (10) and (11) in the supplements of [2]. Equation (10) is replaced by

$$\frac{\gamma_n(t)}{|A(\theta + a_{n, \varepsilon}^{-1} t)|^{1/2}} = \frac{\gamma_n(t)}{|A(\theta)|^{1/2}} + a_{n, \varepsilon}^{-1} \gamma_n(t) D \frac{1}{|A(\theta + e_t)|^{1/2}} t,$$

where $\|e_t\| \leq \delta$, and this leads to replacing $\pi(\theta_0) \int_{\tau(B_\delta) \times \mathbb{R}^d} g_n(t, \nu) dt d\nu$ in equation (11) by

$\int_{\tau(B_\delta) \times \mathbb{R}^d} \gamma_n(t) g_n(t, \nu) dt d\nu$. These changes have no effect on the arguments therein since $\sup_{t \in t(B_\delta)} \gamma_n(t) = O_P(1)$ by Condition 4. Therefore (i) holds.

For (ii), By Condition 5 and Lemma A2 of [1], there exists a $\delta' < \delta$ such that $\inf_{t \in t(B_{\delta'})} \gamma_n(t) = \Theta_P(1)$ and $\int_{\mathbb{R}^d} \int_{t(B_{\delta'})} g_n(t, \nu) dt d\nu = \Theta_P(1)$. Then since $g_{B_\delta, r}(1) \geq \inf_{t \in t(B_{\delta'})} \gamma_n(t) \int_{\mathbb{R}^d} \int_{t(B_{\delta'})} g_n(t, \nu) dt d\nu$, (ii) holds.

For (iii), if $(k_1, k_2) = (1, 0)$ then $\tilde{q}_{B_\delta, tv}(t)/\tilde{q}_{B_\delta, tv}(1)$ can be expanded by following the arguments in the proof of Lemma 5 of [2]. For the other pairs of (k_1, k_2) , $\tilde{q}_{B_\delta, tv}(t^{k_1} v^{k_2})/\tilde{q}_{B_\delta, tv}(1)$, can be expanded similarly as in the proof of Lemma 4 from [1].

For (iv), $\gamma_n(t)$ plays the same role as $\pi(\theta)$ in the proof of Lemma 5 in [2], and the arguments therein can be followed exactly. The term τ_n^p is from the definition of $\gamma_n(t)$ that $r_n(\theta_0 + a_{n,\varepsilon}^{-1}t) = \tau_n^p \gamma_n(t)$. \square

Recall the definition of the estimator $\theta_\varepsilon = \int \theta dQ_\varepsilon(\theta | s_{\text{obs}}) d\theta$. Define the expectation of θ with distribution $\tilde{Q}_\varepsilon(\theta \in A | s_{\text{obs}})$ as $\tilde{\theta}_\varepsilon$ and the expectation of the regression adjusted values, θ^* with density $\tilde{q}_\varepsilon(\theta, s)$ as $\tilde{\theta}_\varepsilon^*$. Let $E_{G,r}(\cdot)$ be the expectation with the density $G_n(v)^{(\text{norm})}$, and $E_{G,r}\{h(v)\}$ can be written as $g_{B_\delta, r}\{h(v)\}/g_{B_\delta, r}(1)$. Let $\psi(\nu) = k_n^{-1} \beta_0 \{A(\theta_0)^{1/2} W_{\text{obs}} + a_n \varepsilon_n \nu\}$, where $k_n = 1$, if $c_\varepsilon < \infty$, and $a_n \varepsilon_n$, if $c_\varepsilon = \infty$.

Lemma 4. Assume Condition 3-6 and 7. Then if $\varepsilon_n = o(a_n^{-1/2})$,

- (i) $\tilde{\theta}_\varepsilon = \theta_0 + a_n^{-1} \beta_0 A(\theta_0)^{1/2} W_{\text{obs}} + \varepsilon_n \beta_0 E_{G_n, r}(\nu) + r_1$, where $r_1 = o_P(a_n^{-1})$;
- (ii) $\tilde{\theta}_\varepsilon^* = \theta_0 + a_n^{-1} \beta_0 A(\theta_0)^{1/2} w_{\text{obs}} + \varepsilon_n (\beta_0 - \beta_\varepsilon) E_{G_n, r}(\nu) + r_2$, where $r_2 = o_P(a_n^{-1})$.

Proof of Lemma 4: These results generalize Lemma A3(c) and Lemma A5(c) in [1] in the sense of permitting use of a data-dependent $r_n(\theta)$, however here we are only considering $\varepsilon_n = o(a_n^{-1/2})$ in contrast to Lemma A5(c) in [1] which assumes the less stringent condition that $\varepsilon_n = o(a_n^{-3/5})$.

With the transformation $t = t(\theta)$, by Lemma 2, if $\varepsilon_n = o(a_n^{-1/2})$,

$$\begin{cases} \tilde{\theta}_\varepsilon = \theta_0 + a_{n,\varepsilon}^{-1} \tilde{q}_{B_\delta, tv}(t)/\tilde{q}_{B_\delta, tv}(1) \\ = \theta_0 + a_{n,\varepsilon}^{-1} g_{B_\delta, r}(t)/g_{B_\delta, r}(1) + o_P(a_n^{-1}), \\ \tilde{\theta}_\varepsilon^* = \theta_0 + a_{n,\varepsilon}^{-1} \tilde{q}_{B_\delta, tv}(t)/\tilde{q}_{B_\delta, tv}(1) - \varepsilon_n \beta_\varepsilon \tilde{q}_{B_\delta, tv}(\nu)/\tilde{q}_{B_\delta, tv}(1) \\ = \theta_0 + a_{n,\varepsilon}^{-1} g_{B_\delta, r}(t)/g_{B_\delta, r}(1) - \varepsilon_n \beta_\varepsilon E_{a_n, r}(\nu) + o_P(a_n^{-1}), \end{cases} \quad (8.4)$$

where the remainder term comes from the fact that $(a_{n,\varepsilon}^{-1} + \varepsilon_n) \{O_P(a_{n,\varepsilon}^{-1}) + O_P(a_n^2 \varepsilon_n^4)\} = o_P(a_n^{-1})$.

First the leading term of $g_{B_\delta, r}(t\nu^k)$ is derived for $k = 0$ or 1. The case of $k = 1$ will be used later. Let $t' = t - \psi(\nu)$, then

$$\begin{aligned} g_{B_\delta, r}(t\nu^k) &= \int_{\mathbb{R}^d} \int_{t(B_\delta)} \{t' + \psi(\nu)\} \nu^k \gamma_n(t) g_n(t, \nu) dt d\nu \\ &= \int_{\mathbb{R}^d} \psi(\nu) \nu^k G_{n, r}(\nu) d\nu \end{aligned}$$

$$+ \int_{\mathbb{R}^d} \int_{t(B_\delta)} t' \nu^k \gamma_n(t) g_n(t, \nu) dt d\nu.$$

By matrix algebra, it is straightforward to show that $g_n(t, \nu) = N\{t; \psi(\nu), k_n^{-2} I(\theta_0)^{-1}\} G_n(\nu)$. Then with the transformation t' , we have

$$\begin{aligned} g_{B_\delta, r}(t\nu^k) &- \int_{\mathbb{R}^d} \psi(\nu) \nu^k G_{n, r}(\nu) d\nu = \\ &\int_{\mathbb{R}^d} \int_{t(B_\delta) - \psi(\nu)} t' \nu^k \gamma_n\{\psi(\nu) + t'\} N\{t'; 0, k_n^{-2} I(\theta_0)^{-1}\} G_n(\nu) dt' d\nu. \end{aligned}$$

By applying the Taylor expansion on $\gamma_n\{\psi(\nu) + t'\}$, the right hand side of the above equation is equal to

$$\begin{aligned} &\int_{\mathbb{R}^d} \int_{t(B_\delta) - \psi(\nu)} t' N\{t'; 0, k_n^{-2} I(\theta_0)^{-1}\} dt' \cdot \gamma_n\{\psi(\nu)\} \nu^k G_n(\nu) d\nu \\ &+ \int_{\mathbb{R}^d} \int_{t(B_\delta) - \psi(\nu)} t'^2 D_t \gamma_n\{\psi(\nu) + e_t\} N\{t'; 0, k_n^{-2} I(\theta_0)^{-1}\} dt' \cdot \nu^k G_n(\nu) d\nu = \\ &k_n^{-1} \int_{\mathbb{R}^d} \int_{Q_v} t'' N\{t''; 0, I(\theta_0)^{-1}\} dt'' \cdot \gamma_n\{\psi(\nu)\} \nu^k G_n(\nu) d\nu \quad (8.5) \\ &+ k_n^{-2} \int_{\mathbb{R}^d} \int_{Q_v} t''^2 D_t \gamma_n\{\psi(\nu) + e_t\} N\{t''; 0, I(\theta_0)^{-1}\} dt'' \cdot \nu^k G_n(\nu) d\nu, \end{aligned}$$

where $Q_v = \{a_n(\theta - \theta_0) - k_n \psi(\nu) | \theta \in B_\delta\}$ and $t'' = k_n t'$. Since Q_v can be written as $\{a_n(\theta - \theta_0 - \beta_0 \varepsilon_n \nu) - \beta_0 A(\theta_0)^{1/2} W_{\text{obs}} | \theta \in B_\delta\}$, it converges to \mathbb{R}^p for any fixed v with probability one using the dominated convergence theorem. Then $\int_{Q_v} t'' N\{t''; 0, \tau(\theta_0)^{-1}\} dt'' = o_P(1)$ for fixed v , and by the continuous mapping theorem and Condition 4, the first term in the right hand side of (8.5) is of the order $o_P(k_n^{-1})$. The second term is bounded by $k_n^{-2} \sup_{t \in \mathbb{R}} \|D_t \gamma_n(t)\| \int_{\mathbb{R}^p} \|t''\|^2 N\{t''; 0, I(\theta_0)^{-1}\} dt'' \int_{\mathbb{R}^d} \nu^k G_n(\nu) d\nu$, which is of the order $O_P(k^{-2} \tau_n / a_{n,\varepsilon})$ by Condition 6. Therefore

$$g_{B_\delta, r}(t\nu^k) = \int_{\mathbb{R}^d} \psi(\nu) \nu^k G_n(\nu) d\nu + o_P(k_n^{-1}). \quad (8.6)$$

By algebra, $k_n = a_{n,\varepsilon}^{-1} a_n$, and

$$\begin{aligned} &\int_{\mathbb{R}^d} \psi(\nu) \nu^k G_n(\nu) d\nu = \\ &a_{n,\varepsilon} \beta_0 \{a_n^{-1} A(\theta_0)^{1/2} W_{\text{obs}} \int_{\mathbb{R}^d} \nu^k G_{n, r}(\nu) d\nu + \varepsilon_n \int_{\mathbb{R}^d} \nu^{k+1} G_{n, r}(\nu) d\nu\}. \end{aligned} \quad (8.7)$$

Then (i) and (ii) in the Lemma holds by plugging the expansion of $g_{B_\delta, r}(t)$ into (8.4). \square

Lemma 5. Assume Condition 3, 4, 2-9. Then as $n \rightarrow \infty$,

- (i) For any $\delta < \delta_0$, $r_{B_\delta^c}(1)$ and $\tilde{q}_{B_\delta^c}(1)$ are $o_P(\tau_n^p)$. More specifically, they are of the order $O_P\left(\tau_n^p e^{-\alpha_\delta a_{n,\varepsilon} c_\delta}\right)$ for some positive constants c_δ and α_δ depending on δ .
- (ii) $q_{B_\delta}(1) = \tilde{q}_{B_\delta}(1) \{1 + O_P(\alpha_n^{-1})\}$ and $\sup_{A \subset B_\delta} |q_A(1) - \tilde{q}_A(1)|/\tilde{q}_{B_\delta}(1) = O_P(\alpha_n^{-1})$;
- (iii) if $\varepsilon_n = o(a_n^{-1/2})$, then $\tilde{q}_{B_\delta}(1)$ and $r_{B_\delta}(1)$ are $\Theta_P(\tau_n^p a_{n,\varepsilon}^{d-p})$, and thus $\tilde{q}_{P_0}(1)$ and $q_{P_0}(1)$ are $\Theta_P(\tau_n^p a_{n,\varepsilon}^{d-p})$;
- (iv) if $\varepsilon_n = o(a_n^{-1/2})$, $\theta_\varepsilon = \tilde{\theta}_\varepsilon + o_P(a_{n,\varepsilon}^{-1})$. If $\varepsilon_n = o(a_n^{-3/5})$, $\theta_\varepsilon = \tilde{\theta}_\varepsilon + o_P(a_n^{-1})$.

Proof of Lemma 5: This generalizes Lemma 2 in the supplements of [1]. The arguments therein can be followed exactly, by Condition 4 and the fact that regarding $\pi(\theta)$, only the condition $\sup_{\theta \in \mathbb{R}^p} \pi(\theta) < \infty$ is used.

Lemma 6. Assume Condition 3, 4, 2-9.

- (i) For any $\delta < \delta_0$, $Q_\varepsilon(\theta \in B_\delta^c \mid s_{\text{obs}})$ and $\tilde{Q}_\varepsilon(\theta \in B_\delta^c \mid s_{\text{obs}})$ are $o_p(1)$;
- (ii) There exists some $\delta < \delta_0$ such that

$$\sup_{A \in \mathfrak{B}^p} |Q_\varepsilon(\theta \in A \cap B_\delta \mid s_{\text{obs}}) - \tilde{Q}_\varepsilon(\theta \in A \cap B_\delta \mid s_{\text{obs}})| = o_p(1);$$

- (iii) $a_{n,\varepsilon}(\theta_\varepsilon - \tilde{\theta}_\varepsilon) = o_p(1)$.

Proof of Lemma 6: This lemma generalizes Lemma A3 of [1]. The proof of Lemma A3 of [1] only needs Lemma 3 and 5 from [2] to hold. The result that $q_{B_\delta^c}\{h(\theta)\} = O_p(\tau_n^p e^{-\alpha_n^\delta c_\delta})$ for some positive constants α_δ and c_δ , which generalizes the case of $r_n(\theta) = \pi(\theta)$ in Lemma 3 of [2], holds by Condition 4, since the latter only uses the fact that $\sup_{\theta \in B_\delta^c} \pi(\theta) < \infty$. Then the arguments in the proof of Lemma 3 in [2] can be followed exactly, despite the term τ_n^p that is not included in the order of $\pi_{B_\delta^c}\{h(\theta)\}$, since $Q_\varepsilon(\theta \in A \mid s_{\text{obs}})$ is the ratio $q_A(1)/q_{\mathbb{R}^p}(1)$. Since Lemma 5 in [2] has been generalized by Lemma (3) above, the arguments of the proof of Lemma A3 in [1] can be followed exactly. \square

With the above lemmas holding for $\varepsilon_n = o(a_n^{-1/2})$, lines for proving Proposition 1 in [1] can be followed exactly to finish the proof of Theorem 2.

9. PROOF OF THEOREM 3

Lemma 7. Assume Condition 3-10. If $\varepsilon_n = o_p(a_n^{-3/5})$, then $a_n \varepsilon_n (\beta_\varepsilon - \beta_0) = o(1)$.

Proof of Lemma 7: This generalizes Lemma A4 in [1] by replacing $\pi(\theta_0 + a_{n,\varepsilon}^{-1}t)$ therein with $\gamma_n(t)$. By Condition 4 and the arguments in the proof of Lemma A4 in [1], it can be shown that

$$\frac{q_{\mathbb{R}^p}\{(\theta - \theta_0)^{k_1}(s - s_{\text{obs}})^{k_2}\}}{q_{\mathbb{R}^p}(1)} = a_{n,\varepsilon}^{-k_1} \varepsilon_n^{-k_2} \left\{ \frac{\tilde{q}_{B_\delta, tv}(t^{k_1} \nu^{k_2})}{\tilde{q}_{B_\delta, tv}(1)} + O_p(\alpha_n^{-1}) \right\}.$$

Then by Lemma 2 (iii), the right hand side of the above is equal to

$$a_{n,\varepsilon}^{-k_1} \varepsilon_n^{-k_2} \left\{ \frac{g_{B_\delta, r}(t^{k_1} \nu^{k_2})}{g_{B_\delta, r}(1)} + O_p(a_{n,\varepsilon}^{-1}) + O_p(a_n^2 \varepsilon_n^4) + O_p(\alpha_n^{-1}) \right\}.$$

Since $\beta_\varepsilon = \text{Cov}_\varepsilon(\theta, S_n) \text{Var}_\varepsilon(S_n)^{-1}$,

$$a_n \varepsilon_n (\beta_\varepsilon - \beta_0) = k_n \left[\frac{g_{B_\delta, r}(t \nu)}{g_{B_\delta, r}(1)} - \frac{g_{B_\delta, r}(t) g_{B_\delta, r}(\nu)}{g_{B_\delta, r}(1)^2} + o_p(k_n^{-1}) \right] \times$$

$$\left[\frac{g_{B_\delta, r}(\nu \nu^T)}{g_{B_\delta, r}(1)} - \frac{g_{B_\delta, r}(\nu) g_{B_\delta, r}(\nu)^T}{g_{B_\delta, r}(1)^2} + o_p(k_n^{-1}) \right]^{-1} - a_n \varepsilon_n \beta_0,$$

where the equations that $a_{n,\varepsilon}^{-1} k_n = o(1)$, $a_n^2 \varepsilon_n^4 k_n = o(p)$, and $\alpha_n^{-1} k_n = o(a_n^{-2/5} k_n) = o(1)$ are used. By algebra, the right hand side of the equation above can be rewritten as

$$\left\{ \frac{g_{B_\delta, r}(\{k_n t - a_n \varepsilon_n \beta_0 \nu\})}{g_{B_\delta, r}(1)} - \frac{g_{B_\delta, r}(k_n t - a_n \varepsilon_n \beta_0 \nu) g_{B_\delta, r}(\nu)}{g_{B_\delta, r}(1)^2} + o_p(1) \right\} \times \left\{ E_{G, r}(\nu \nu^T) - E_{G, r}(\nu) E_{G, r}(\nu)^T + o_p(k_n^{-1}) \right\}^{-1}.$$

By plugging (8.6) and (8.7) in the above, $a_n \varepsilon_n (\beta_\varepsilon - \beta_0)$ is equal to

$$\begin{aligned} & \{E_{G, r}(\nu) \beta_0 A(\theta_0)^{1/2} W_{\text{obs}} - E_{G, r}(\nu) \beta_0 A(\theta_0)^{1/2} W_{\text{obs}} + o_p(1)\} \times \\ & \{\text{Var}_{G, r}(\nu) + o_p(k_n^{-1})\}^{-1} \\ & = o_p(1) \{\text{Var}_{G, r}(\nu) + o_p(k_n^{-1})\}^{-1}. \end{aligned}$$

Since

$$\text{Var}_{G, r}(\nu) \geq \frac{\inf_{t \in t(B_{\delta'})} \gamma_n(t)}{g_{B_\delta, r}(1)} \int_{\mathbb{R}^d} \int_{t(B_{\delta'})} \{\nu - E_{G, r}(\nu)\}^2 g_n(t, \nu) dt d\nu,$$

where δ' is defined in the proof of Lemma 3(ii), we have $\text{Var}_{G, r}(\nu)^{-1} = \Theta_p(1)$. Therefore $a_n \varepsilon_n (\beta_\varepsilon - \beta_0) = o_p(1)$. \square

Lemma 8. Results generalizing Lemma A5 in [1], i.e. replacing Π_ε and $\tilde{\Pi}_\varepsilon$ therein with Q_ε and \tilde{Q}_ε , hold.

Proof of Lemma 8: In [1], the proof of Lemma A5 in [1] requires Lemma A4 and Lemma 2 in the supplements therein to hold. Since we have show that their generalized results hold for $\varepsilon_n = o_p(a_n^{-1/2})$, (see Lemma 7 and Lemma 5 above), the proof of this lemma for $\varepsilon_n = o_p(a_n^{-3/5})$ follows the same arguments in [1], replacing $\pi(\theta)$ with a $r_n(\theta)$ that satisfies conditions 3 - 6. \square

With all above lemmas, the proof of Theorem 3 holds by following the same arguments as those in the proof of Theorem 1 in [1].

REFERENCES

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