

1. Asymptotic normality

Theorem 1.1. *If $h^{-1}n^{-1/2}a_n \xrightarrow{p} 0$ and $hn^{-1/2}b_n \xrightarrow{p} \infty$ then $\hat{\beta}(s) - \beta(s) - \frac{\kappa_2 h^2}{2\kappa_0} \{\beta_{uu}(s) + \beta_{vv}(s)\} = O_p(n^{-1/2}h^{-1})$*

Proof. The idea of the proof is to show that the objective being minimized achieves a unique minimum, which must be $\hat{\beta}(s)$.

The order of convergence is $hn^{1/2}$ where $h = O(n^{-1/6})$.

We find the limiting distribution of the estimator:

$$\begin{aligned}
V_4^{(n)}(\mathbf{u}) &= Q \left\{ \beta(s) + h^{-1}n^{-1/2}\mathbf{u} \right\} - Q \left\{ \beta(s) \right\} \\
&= (1/2) \left[\mathbf{Y} - \mathbf{Z}(s) \left\{ \beta(s) + h^{-1}n^{-1/2}\mathbf{u} \right\} \right]^T \mathbf{W}(s) \left[\mathbf{Y} - \mathbf{Z}(s) \left\{ \beta(s) + h^{-1}n^{-1/2}\mathbf{u} \right\} \right] \\
&\quad + \sum_{j=1}^p \lambda_j \left\| \beta(s) + h^{-1}n^{-1/2}\mathbf{u}_j \right\| \\
&\quad - (1/2) \left\{ \mathbf{Y} - \mathbf{Z}(s)\beta(s) \right\}^T \mathbf{W}(s) \left\{ \mathbf{Y} - \mathbf{Z}(s)\beta(s) \right\} - \sum_{j=1}^p \lambda_j \left\| \beta(s) \right\| \\
&= (1/2) \mathbf{u}^T \left\{ h^{-2}n^{-1} \mathbf{Z}^T(s) \mathbf{W}(s) \mathbf{Z}(s) \right\} \mathbf{u} - \mathbf{u}^T \left[h^{-1}n^{-1/2} \mathbf{Z}^T(s) \mathbf{W}(s) \left\{ \mathbf{Y} - \mathbf{Z}(s)\beta(s) \right\} \right] \\
&\quad + \sum_{j=1}^p n^{-1/2} \lambda_j n^{1/2} \left\{ \left\| \beta_j(s) + h^{-1}n^{-1/2}\mathbf{u}_j \right\| - \left\| \beta_j(s) \right\| \right\} \tag{1}
\end{aligned}$$

Note the different limiting behavior of the third term between the cases $j \leq p_0$ and $j > p_0$:

Case $j \leq p_0$. If $j \leq p_0$ then $n^{-1/2}\lambda_j \rightarrow n^{-1/2}\lambda \|\beta_j(s)\|^{-\gamma}$ and $|\sqrt{n} \{ \|\beta_j(s) + h^{-1}n^{-1/2}\mathbf{u}_j\| - \|\beta_j(s)\| \}| \leq h^{-1}\|\mathbf{u}_j\|$ so $\lim_{n \rightarrow \infty} \lambda_j (\|\beta_j(s) + h^{-1}n^{-1/2}\mathbf{u}_j\| - \|\beta_j(s)\|) \leq h^{-1}n^{-1/2}\lambda_j \|\mathbf{u}_j\| \leq h^{-1}n^{-1/2}a_n \|\mathbf{u}_j\| \rightarrow 0$

Case $j > p_0$. If $j > p_0$ then $\lambda_j (\|\beta_j(s) + h^{-1}n^{-1/2}\mathbf{u}_j\| - \|\beta_j(s)\|) = \lambda_j h^{-1}n^{-1/2}\|\mathbf{u}_j\|$.

And note that $h = O(n^{-1/6})$ so that if $hn^{-1/2}b_n \xrightarrow{p} \infty$ then $h^{-1}n^{-1/2}b_n \xrightarrow{p} \infty$.

Now, if $\|\mathbf{u}_j\| \neq 0$ then $h^{-1}n^{-1/2}\lambda_j\|\mathbf{u}_j\| \geq h^{-1}n^{-1/2}b_n\|\mathbf{u}_j\| \rightarrow \infty$. On the other hand, if $\|\mathbf{u}_j\| = 0$ then $h^{-1}n^{-1/2}\lambda_j\|\mathbf{u}_j\| = 0$.

Thus, the limit of $V_4^{(n)}(\mathbf{u})$ is the same as the limit of $V_4^{*(n)}(\mathbf{u})$ where

$$V_4^{*(n)}(\mathbf{u}) = \begin{cases} (1/2)\mathbf{u}^T \{h^{-2}n^{-1}\mathbf{Z}^T(\mathbf{s})\mathbf{W}(\mathbf{s})\mathbf{Z}(\mathbf{s})\} \mathbf{u} - \mathbf{u}^T [h^{-1}n^{-1/2}\mathbf{Z}^T(\mathbf{s})\mathbf{W}(\mathbf{s})\{\mathbf{Y} - \mathbf{Z}(\mathbf{s})\beta(\mathbf{s})\}] & \text{if } \|\mathbf{u}_j\| = 0 \ \forall j > p_0 \\ \infty & \text{otherwise} \end{cases}$$

And $V_4^{*(n)}(\mathbf{u})$ is convex and is minimized at $\hat{\mathbf{u}}^{(n)}$:

$$\begin{aligned} 0 &= \{h^{-2}n^{-1}\mathbf{Z}^T(\mathbf{s})\mathbf{W}(\mathbf{s})\mathbf{Z}(\mathbf{s})\} \hat{\mathbf{u}}^{(n)} - [h^{-1}n^{-1/2}\mathbf{Z}^T(\mathbf{s})\mathbf{W}(\mathbf{s})\{\mathbf{Y} - \mathbf{Z}(\mathbf{s})\beta(\mathbf{s})\}] \\ \therefore \hat{\mathbf{u}}^{(n)} &= \{n^{-1}\mathbf{Z}^T(\mathbf{s})\mathbf{W}(\mathbf{s})\mathbf{Z}(\mathbf{s})\}^{-1} [hn^{-1/2}\mathbf{Z}^T(\mathbf{s})\mathbf{W}(\mathbf{s})\{\mathbf{Y} - \mathbf{Z}(\mathbf{s})\beta(\mathbf{s})\}] \end{aligned} \tag{2}$$

By the epiconvergence results of ? and ?, the minimizer of the limiting function is the limit of the minimizers $\hat{\mathbf{u}}^{(n)}$. And since, by Lemma 2 of ?,

$$\hat{\mathbf{u}}^{(n)} \xrightarrow{d} N(0, f(\mathbf{s})\kappa_0^{-2}\nu_0\sigma^2\Psi^{-1}) \tag{3}$$

the result is proven. □

2. Selection

Theorem 2.1. *If $hn^{-1/2}b_n \xrightarrow{p} \infty$ then $P\left\{\hat{\beta}_{(b)}(s) = 0\right\} \rightarrow 1$.*

Proof. We showed in Theorem 1.1 that the

The proof is by contradiction.

Recall that the objective to be minimized by $\hat{\beta}_{(p)}(s)$ is

$$Q\left\{\hat{\beta}(s)\right\} = (1/2)\left\{\mathbf{Y} - \mathbf{Z}(s)\hat{\beta}(s)\right\}^T \mathbf{W}(s)\left\{\mathbf{Y} - \mathbf{Z}(s)\hat{\beta}(s)\right\} + \sum_{j=1}^p \lambda_j \|\hat{\beta}_p(s)\| \quad (4)$$

Let $\hat{\beta}_p(s) \neq 0$. Then (4) is differentiable w.r.t. $\beta_p(s)$ and Q is maximized at

$$\begin{aligned} 0 &= \mathbf{Z}_{(p)}^T(s) \mathbf{W}(s) \left\{ \mathbf{Y} - \mathbf{Z}_{(-p)}(s) \hat{\beta}_{(-p)}(s) - \mathbf{Z}_{(p)}(s) \hat{\beta}_{(p)}(s) \right\} - \lambda_p \frac{\hat{\beta}_{(p)}(s)}{\|\hat{\beta}_{(p)}(s)\|} \\ &= \mathbf{Z}_{(p)}^T(s) \mathbf{W}(s) \left[\mathbf{Y} - \mathbf{Z}(s) \beta(s) - \frac{h^2 \kappa_2}{2\kappa_0} \{\beta_{uu}(s) + \beta_{vv}(s)\} \right] \\ &\quad + \mathbf{Z}_{(p)}^T(s) \mathbf{W}(s) \mathbf{Z}_{(-p)}(s) \left[\beta_{(-p)}(s) + \frac{h^2 \kappa_2}{2\kappa_0} \{\beta_{(-p),uu}(s) + \beta_{(-p),vv}(s)\} - \hat{\beta}_{(-p)}(s) \right] \\ &\quad + \mathbf{Z}_{(p)}^T(s) \mathbf{W}(s) \mathbf{Z}_{(p)}(s) \left[\beta_{(p)}(s) + \frac{h^2 \kappa_2}{2\kappa_0} \{\beta_{(p),uu}(s) + \beta_{(p),vv}(s)\} - \hat{\beta}_{(p)}(s) \right] \\ &\quad - \lambda_p \frac{\hat{\beta}_{(p)}(s)}{\|\hat{\beta}_{(p)}(s)\|} \end{aligned} \quad (5)$$

So

$$\begin{aligned}
\frac{h}{\sqrt{n}} \lambda_p \frac{\hat{\beta}_{(p)}(\mathbf{s})}{\|\hat{\beta}_{(p)}(\mathbf{s})\|} &= \mathbf{Z}_{(p)}^T(\mathbf{s}) \mathbf{W}(\mathbf{s}) \frac{h}{\sqrt{n}} \left[\mathbf{Y} - \mathbf{Z}(\mathbf{s}) \boldsymbol{\beta}(\mathbf{s}) - \frac{h^2 \kappa_2}{2\kappa_0} \{\beta_{uu}(\mathbf{s}) + \beta_{vv}(\mathbf{s})\} \right] \\
&+ \left\{ n^{-1} \mathbf{Z}_{(p)}^T(\mathbf{s}) \mathbf{W}(\mathbf{s}) \mathbf{Z}_{(-p)}(\mathbf{s}) \right\} h \sqrt{n} \left[\boldsymbol{\beta}_{(-p)}(\mathbf{s}) + \frac{h^2 \kappa_2}{2\kappa_0} \{\beta_{(-p),uu}(\mathbf{s}) + \beta_{(-p),vv}(\mathbf{s})\} - \hat{\boldsymbol{\beta}}_{(-p)}(\mathbf{s}) \right] \\
&+ \left\{ n^{-1} \mathbf{Z}_{(p)}^T(\mathbf{s}) \mathbf{W}(\mathbf{s}) \mathbf{Z}_{(-p)}(\mathbf{s}) \right\} h \sqrt{n} \left[\boldsymbol{\beta}_{(-p)}(\mathbf{s}) + \frac{h^2 \kappa_2}{2\kappa_0} \{\beta_{(p),uu}(\mathbf{s}) + \beta_{(p),vv}(\mathbf{s})\} - \hat{\boldsymbol{\beta}}_{(p)}(\mathbf{s}) \right]
\end{aligned} \tag{6}$$

From Lemma 2 of ?, $\left\{ n^{-1} \mathbf{Z}_{(p)}^T(\mathbf{s}) \mathbf{W}(\mathbf{s}) \mathbf{Z}_{(-p)}(\mathbf{s}) \right\} = O_p(1)$ and $\left\{ n^{-1} \mathbf{Z}_{(p)}^T(\mathbf{s}) \mathbf{W}(\mathbf{s}) \mathbf{Z}_{(p)}(\mathbf{s}) \right\} = O_p(1)$.

From Theorem 3 of ?, we have that $h \sqrt{n} \left[\hat{\boldsymbol{\beta}}_{(-p)}(\mathbf{s}) - \boldsymbol{\beta}_{(-p)}(\mathbf{s}) - \frac{h^2 \kappa_2}{2\kappa_0} \{\beta_{(-p),uu}(\mathbf{s}) + \beta_{(-p),vv}(\mathbf{s})\} \right] = O_p(1)$ and $h \sqrt{n} \left[\hat{\boldsymbol{\beta}}_{(p)}(\mathbf{s}) - \boldsymbol{\beta}_{(p)}(\mathbf{s}) - \frac{h^2 \kappa_2}{2\kappa_0} \{\beta_{(p),uu}(\mathbf{s}) + \beta_{(p),vv}(\mathbf{s})\} \right] = O_p(1)$.

So the second and third terms of the sum in (6) are $O_p(1)$.

We showed in the proof of 1.1 that $h \sqrt{n} \mathbf{Z}_{(p)}^T(\mathbf{s}) \mathbf{W}(\mathbf{s}) \left[\mathbf{Y} - \mathbf{Z}(\mathbf{s}) \boldsymbol{\beta}(\mathbf{s}) - \frac{h^2 \kappa_2}{2\kappa_0} \{\beta_{uu}(\mathbf{s}) + \beta_{vv}(\mathbf{s})\} \right] = O_p(1)$.

The three terms of the sum to the right of the equals sign in (6) are $O_p(1)$, so for $\hat{\boldsymbol{\beta}}_{(p)}(\mathbf{s})$ to be a solution, we must have that $h n^{-1/2} \lambda_p \frac{\hat{\beta}_{(p)}(\mathbf{s})}{\|\hat{\beta}_{(p)}(\mathbf{s})\|} = O_p(1)$.

But since by assumption $\hat{\beta}_{(p)}(\mathbf{s}_i) \neq 0$, there must be some $k \in \{1, \dots, d_p\}$ such that $|\hat{\beta}_{(p),k}(\mathbf{s})| = \max\{|\hat{\beta}_{(p),k'}(\mathbf{s})| : 1 \leq k' \leq d_p\}$. And for this k , we have that $|\hat{\beta}_{(p),k}(\mathbf{s})| / \|\hat{\boldsymbol{\beta}}_{(p)}(\mathbf{s})\| \geq 1/\sqrt{d_p} > 0$.

Now since $h n^{-1/2} b_n \rightarrow \infty$, we have that $h n^{-1/2} \lambda_p \frac{\hat{\beta}_{(p)}(\mathbf{s})}{\|\hat{\beta}_{(p)}(\mathbf{s})\|} \geq h n^{-1/2} b_n d_p^{-1/2} \rightarrow \infty$ and therefore the term to the left of the equals sign dominates the sum to the right of the equals sign in (6). So for large enough n , $\hat{\boldsymbol{\beta}}_{(p)}(\mathbf{s}) \neq 0$ cannot maximize Q .

So $P \left\{ \hat{\boldsymbol{\beta}}_{(b)}(\mathbf{s}) = 0 \right\} \rightarrow 1$. □

3. A note on rates

To prove the oracle properties of LAGR, we assumed that $h^{-1}n^{-1/2}a_n \xrightarrow{p} 0$ and $hn^{-1/2}b_n \xrightarrow{p} \infty$.

Therefore, $h^{-1}n^{-1/2}\lambda_n \rightarrow 0$ for $j \leq p_0$ and $hn^{-1/2}\lambda_n\|\beta_j(\mathbf{s})\|^{-\gamma} \rightarrow \infty$ for $j > p_0$.

We require a λ_n that can satisfy both assumptions. Suppose $\lambda_n = n^\alpha$, and recall that $h = O(n^{-1/6})$.

Then $h^{-1}n^{-1/2}\lambda_n = O(n^{-1/3+\alpha})$ and $hn^{-1/2}\lambda_n(h\sqrt{n})^\gamma = O(n^{-2/3+\alpha+\gamma/3})$.

So $(2 - \gamma)/3 < \alpha < 1/3$, which can only be satisfied for $\gamma > 1$.