## Web-based Supplementary Material for "Local Adaptive Grouped Regularization and its Oracle Properties for Varying Coefficient Regression"

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## 1. Lemmas

## Lemma 1.

$$E\left[\sum_{i=1}^{n} q_{1}\left(\boldsymbol{Z}_{i}^{T}\boldsymbol{\zeta}(\boldsymbol{s}), Y_{i}\right) \boldsymbol{Z}_{i} K_{h}\left(\|\boldsymbol{s}-\boldsymbol{s}_{i}\|\right)\right] = \left(2^{-1} n^{1/2} h^{3} f(\boldsymbol{s}) \kappa_{2} \boldsymbol{\Gamma}(\boldsymbol{s}) \left(\nabla_{uu}^{2} \boldsymbol{\beta}(\boldsymbol{s}) + \nabla_{vv}^{2} \boldsymbol{\beta}(\boldsymbol{s})\right)^{T}\right) + o_{p}\left(h^{2} \mathbf{1}_{3p}\right)$$

$$\mathbf{0}_{2p}$$

and

$$Var\left[\sum_{i=1}^{n} q_1 \left(\mathbf{Z}_i^T \boldsymbol{\zeta}(\boldsymbol{s}), Y_i\right) \mathbf{Z}_i K_h \left(\|\boldsymbol{s} - \boldsymbol{s}_i\|\right)\right] = f(\boldsymbol{s}) \operatorname{diag}\left\{\nu_0, \nu_2, \nu_2\right\} \otimes \boldsymbol{\Gamma}(\boldsymbol{s}) + o\left(1\right)$$
$$= \Lambda + o\left(1\right)$$

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*Proof.* Expectation: For j = 1, ..., p, by a Taylor expansion of  $\beta_j(s_i)$  around s,

$$eta_j(oldsymbol{s}_i) = eta_j(oldsymbol{s}) + 
abla eta_j(oldsymbol{s})(oldsymbol{s}_i - oldsymbol{s}) + (oldsymbol{s}_i - oldsymbol{s})^T \left\{ 
abla^2 eta_j(oldsymbol{s}) \right\} (oldsymbol{s}_i - oldsymbol{s}) + o\left(h^2\right)$$

and thus, for  $\boldsymbol{x} \in \mathbb{R}^p$ ,

$$\boldsymbol{x}_i^T \boldsymbol{\beta}(\boldsymbol{s}_i) = \sum_{j=1}^p x_{ij} \left[ \beta_j(\boldsymbol{s}) + \nabla \beta_j(\boldsymbol{s})^T (\boldsymbol{s}_i - \boldsymbol{s}) + \tilde{\beta}_{ij}'' \right] + o(h^2).$$

Letting  $\boldsymbol{z}_i^T = \{(1, s_{i,1} - s_1, s_{i,2} - s_2) \otimes \boldsymbol{x}_i^T\}$  and  $\boldsymbol{\zeta}(\boldsymbol{s}) = (\boldsymbol{\beta}(\boldsymbol{s})^T, \nabla_u \boldsymbol{\beta}(\boldsymbol{s})^T, \nabla_v \boldsymbol{\beta}(\boldsymbol{s})^T)^T$ , we have that

$$egin{aligned} oldsymbol{x}_i^T oldsymbol{eta}(oldsymbol{s}_i) - oldsymbol{z}_i^T oldsymbol{\zeta}(oldsymbol{s}) &= & x_i^T oldsymbol{ ilde{eta}}_i'' + o\left(h^2
ight) \ &= & O_p\left(h^2
ight). \end{aligned}$$

By a Taylor expansion around  $\boldsymbol{x}^T\boldsymbol{\beta}(\boldsymbol{s}_i)$ , then,

$$q_1\left(\boldsymbol{z}_i^T \boldsymbol{\zeta}(\boldsymbol{s}), \mu(\boldsymbol{s}_i, \boldsymbol{z}_i)\right) = q_1\left(\boldsymbol{x}_i^T \boldsymbol{\beta}(\boldsymbol{s}_i), \mu(\boldsymbol{s}_i, \boldsymbol{z})\right) - q_2\left(\boldsymbol{x}_i^T \boldsymbol{\beta}(\boldsymbol{s}_i), \mu(\boldsymbol{s}_i, \boldsymbol{z})\right) \boldsymbol{x}_i^T \tilde{\boldsymbol{\beta}}_i'' + o\left(h^2\right).$$

And by the definitions of  $q_1(\cdot)$  and  $q_2(\cdot)$ , we have that

$$q_1\left(\boldsymbol{z}_i^T\boldsymbol{\zeta}(\boldsymbol{s}), \mu(\boldsymbol{s}_i, \boldsymbol{z}_i)\right) = \rho(\boldsymbol{s}_i, \boldsymbol{z}_i)\boldsymbol{x}_i^T\tilde{\boldsymbol{\beta}}_i'' + o\left(h^2\right).$$

Now the expectation of  $\Omega_n$  is

$$nE\left(\omega_{i}|\boldsymbol{Z}_{i}=\boldsymbol{z}_{i},\boldsymbol{s}_{i}\right)=\left(1/2\right)\alpha_{n}\boldsymbol{z}_{i}q_{1}\left(\boldsymbol{z}_{i}^{T}\boldsymbol{\zeta}(\boldsymbol{s}),\mu(\boldsymbol{s}_{i},\boldsymbol{z}_{i})\right)K\left(h^{-1}\|\boldsymbol{s}-\boldsymbol{s}_{i}\|\right)$$

$$=\left(1/2\right)\alpha_{n}h^{2}\boldsymbol{z}_{i}\left\{h^{-2}\rho(\boldsymbol{s}_{i},\boldsymbol{z}_{i})\boldsymbol{x}_{i}^{T}\tilde{\boldsymbol{\beta}}_{i}^{"}+o\left(\mathbf{1}_{3p}\right)\right\}K\left(h^{-1}\|\boldsymbol{s}-\boldsymbol{s}_{i}\|\right).$$

To facilitate a change of variables, we observe that  $h^{-2}\tilde{\beta}_j'' = \left(\frac{s_i - s}{h}\right)^T \left\{\nabla^2 \beta_j(s)\right\} \left(\frac{s_i - s}{h}\right)$ . Thus,

$$E\left(\omega_{i}|\boldsymbol{s}_{i}\right)=\left(1/2\right)\alpha_{n}h^{2}\left[\left(\begin{array}{c}1\\h^{-1}(s_{i,1}-s_{1})\\h^{-1}(s_{i,2}-s_{2})\end{array}\right)\otimes\left\{\boldsymbol{\Gamma}(\boldsymbol{s}_{i})h^{-2}\tilde{\boldsymbol{\beta}}_{i}^{\prime\prime}\right\}+o\left(\mathbf{1}_{3p}\right)\right]K\left(h^{-1}\|\boldsymbol{s}-\boldsymbol{s}_{i}\|\right).$$

And, using the symmetry of the kernel function,

$$E(\omega_i) = (1/2)\alpha_n h^4 f(s) \begin{pmatrix} \kappa_2 \\ h\kappa_3 \\ h\kappa_3 \end{pmatrix} \otimes \left[ \mathbf{\Gamma}(s) \left\{ \nabla_{uu}^2 \boldsymbol{\beta}(s) + \nabla_{vv}^2 \boldsymbol{\beta}(s) \right\} \right] + o\left(h^2 \mathbf{1}_{3p}\right)$$

where  $\{\nabla_{uu}^2 \boldsymbol{\beta}(\boldsymbol{s}) + \nabla_{vv}^2 \boldsymbol{\beta}(\boldsymbol{s})\} = (\nabla_{uu}^2 \beta_1(\boldsymbol{s}) + \nabla_{vv}^2 \beta_1(\boldsymbol{s}), \dots, \nabla_{uu}^2 \beta_p(\boldsymbol{s}) + \nabla_{vv}^2 \beta_p(\boldsymbol{s}))^T$ . Thus,

$$E\left(\Omega_{n}\right) = \begin{pmatrix} \alpha_{n}^{-1} 2^{-1} h^{2} \kappa_{2} f(\boldsymbol{s}) \Gamma(\boldsymbol{s}) \left(\nabla_{uu}^{2} \boldsymbol{\beta}(\boldsymbol{s}) + \nabla_{vv}^{2} \boldsymbol{\beta}(\boldsymbol{s})\right)^{T} \\ \mathbf{0}_{2p} \end{pmatrix} + o_{p} \left(h^{2} \mathbf{1}_{3p}\right)$$

**Variance**: By the previous result,  $E(\Omega_n) = O(h^2)$ . Thus,  $var(\Omega_n) \to E(\Omega_n^2)$ , and since the observations are independent,  $E(\Omega_n^2) = \sum_{i=1}^n E(\omega_i^2)$ . And, by Taylor expansion around  $\mathbf{z}_i^T \boldsymbol{\zeta}(\mathbf{s}_i)$ ,

$$q_1^2 \left( \boldsymbol{z}_i^T \boldsymbol{\zeta}(\boldsymbol{s}), Y_i \right) = q_1^2 \left( \boldsymbol{z}_i^T \boldsymbol{\zeta}(\boldsymbol{s}_i), Y_i \right)$$

$$- q_1 \left( \boldsymbol{z}_i^T \boldsymbol{\zeta}(\boldsymbol{s}_i), Y_i \right) q_2 \left( \boldsymbol{z}_i^T \boldsymbol{\zeta}(\boldsymbol{s}_i), Y_i \right) \boldsymbol{x}_i^T \tilde{\boldsymbol{\beta}}_i''$$

$$+ o \left( h^2 \right).$$

Since  $q_1(\cdot,\cdot)$  is the quasi-score function, it follows that

$$E\left(\omega_i^2|\mathbf{Z}_i=\mathbf{z}_i,\mathbf{s}_i\right) = \alpha_n^2 \rho(\mathbf{s}_i,\mathbf{z}_i)\mathbf{z}_i\mathbf{z}_i^T K\left(h^{-1}\|\mathbf{s}-\mathbf{s}_i\|\right) + o\left(h^2\right).$$

By the symmetry of the kernel function,

$$E(\omega_i^2) = n^{-1} f(\mathbf{s}) \operatorname{diag} \{\nu_0, \nu_2, \nu_2\} \otimes \Gamma(\mathbf{s}) + o(1).$$

Thus,

$$Var(\Omega_n) = f(\mathbf{s}) \operatorname{diag} \{\nu_0, \nu_2, \nu_2\} \otimes \Gamma(\mathbf{s}) + o(1).$$

Lemma 2.

$$E\left[\sum_{i=1}^{n} q_{2}\left(\boldsymbol{Z}_{i}^{T}\boldsymbol{\zeta}(\boldsymbol{s}), Y_{i}\right) \boldsymbol{Z}_{i} \boldsymbol{Z}_{i}^{T} K_{h}\left(\|\boldsymbol{s} - \boldsymbol{s}_{i}\|\right)\right] = -f(\boldsymbol{s}) \operatorname{diag}\left\{\kappa_{0}, \kappa_{2}, \kappa_{2}\right\} \otimes \boldsymbol{\Gamma}(\boldsymbol{s}) + o\left(1\right)$$

$$= -\Delta + o\left(1\right)$$

and

$$Var\left\{ \left( \sum_{i=1}^{n} q_2 \left( \mathbf{Z}_i^T \boldsymbol{\zeta}(\boldsymbol{s}), Y_i \right) \mathbf{Z}_i \mathbf{Z}_i^T K_h \left( \| \boldsymbol{s} - \boldsymbol{s}_i \| \right) \right)_{ij} \right\} = O\left( n^{-1} h^{-2} \right)$$

*Proof.* Expectation: The approach is similar to the proof of Lemma 1. By the Taylor expansion of  $q_2\left(\boldsymbol{z}_i^T\boldsymbol{\zeta}(\boldsymbol{s}), \mu\left(\boldsymbol{s}_i, \boldsymbol{z}_i\right)\right)$  around  $\boldsymbol{z}_i^T\boldsymbol{\zeta}(\boldsymbol{s}_i)$ :

$$q_2\left(\boldsymbol{z}_i^T\boldsymbol{\zeta}(\boldsymbol{s}), \mu(\boldsymbol{s}_i, \boldsymbol{z}_i)\right) = q_2\left(\boldsymbol{z}_i^T\boldsymbol{\zeta}(\boldsymbol{s}_i), \mu(\boldsymbol{s}_i, \boldsymbol{z}_i)\right) + q_3\left(\boldsymbol{z}_i^T\boldsymbol{\zeta}(\boldsymbol{s}_i), \mu(\boldsymbol{s}_i, \boldsymbol{z}_i)\right) \left\{\boldsymbol{z}_i^T\boldsymbol{\zeta}(\boldsymbol{s}) - \boldsymbol{z}_i^T\boldsymbol{\zeta}(\boldsymbol{s}_i)\right\}$$
$$= -\rho(\boldsymbol{s}_i, \boldsymbol{z}_i) + o(1).$$

And by the same arguments as before

$$E\left(\delta_{i}|\mathbf{Z}_{i}=\mathbf{z}_{i},\mathbf{s}_{i}\right)=-\alpha_{n}^{2}\rho(\mathbf{s}_{i},\mathbf{z}_{i})\mathbf{z}_{i}\mathbf{z}_{i}^{T}K\left(h^{-1}\|\mathbf{s}_{i}-\mathbf{s}\|\right)$$

$$E\left(\delta_{i}|\mathbf{s}_{i}\right)=-\alpha_{n}^{2}\begin{pmatrix}1\\h^{-1}(s_{i,1}-s_{1})\\h^{-1}(s_{i,2}-s_{2})\end{pmatrix}\begin{pmatrix}1\\h^{-1}(s_{i,1}-s_{1})\\h^{-1}(s_{i,2}-s_{2})\end{pmatrix}^{T}\otimes\mathbf{\Gamma}(\mathbf{s}_{i})K\left(h^{-1}\|\mathbf{s}_{i}-\mathbf{s}\|\right)$$

$$E\left(\delta_{i}\right)=-nf\left(\mathbf{s}\right)\operatorname{diag}\left\{\kappa_{0},\kappa_{2},\kappa_{2}\right\}\otimes\mathbf{\Gamma}\left(\mathbf{s}\right)+o\left(n^{-1}\right)$$

Thus,

$$E(\Delta_n) = -f(s)\operatorname{diag}\left\{\kappa_0, \kappa_2, \kappa_2\right\} \otimes \Gamma(s) + o(1)$$

**Variance**: From the previous result, it follows that  $\{E(\delta_i)\}^2 = O(n^{-2})$ . By the definition of  $\delta_i$ ,

$$E\left(\delta_{i}^{2}|\mathbf{Z}_{i}=\mathbf{z}_{i},\mathbf{s}_{i}\right) = \alpha_{n}^{4}\mathbf{z}_{i}^{T}\mathbf{z}_{i}q_{2}^{2}(\mathbf{s}_{i},\mathbf{z}_{i}) \begin{pmatrix} 1 \\ h^{-1}(s_{i,1}-s_{1}) \\ h^{-1}(s_{i,2}-s_{2}) \end{pmatrix} \begin{pmatrix} 1 \\ h^{-1}(s_{i,1}-s_{1}) \\ h^{-1}(s_{i,2}-s_{2}) \end{pmatrix}^{T} \mathbf{z}_{i}\mathbf{z}_{i}^{T}K^{2}\left(h^{-1}||\mathbf{s}_{i}-\mathbf{s}||\right) + o\left(1\right)$$

And it follows that  $E\left(\delta_{i}^{2}\right)=O\left(n^{-1}\alpha_{n}^{2}\right)$ , and  $Var\left(\Delta_{n}\right)=O\left(\alpha_{n}^{2}\right)$ .