

CONSISTENT VARIABLE SELECTION IN ADDITIVE MODELS

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Supplementary Material

S1. Assumptions

Let $\{(\mathbf{X}_i, Y_i)\}_{i=1}^n$ denote independent pairs, each having the same distribution as (\mathbf{X}, Y) . The technical assumptions we need are as follows.

- (A1) *The density function $f(\mathbf{x})$ of \mathbf{X} is absolutely continuous and compactly supported. Without loss of generality, let its support $\mathcal{X} = [0, 1]^d$. Also there exists constants $0 < c_1 \leq c_2$, such that $c_1 \leq f(\mathbf{x}) \leq c_2$, for all $\mathbf{x} \in \mathcal{X}$.*
- (A2) *The d sets of knots denoted as $k_l = \{0 = x_{l,0} \leq x_{l,1} \leq \dots \leq x_{l,N_l} \leq x_{l,N_l+1} = 1\}$, $l = 1, \dots, d$, are quasi-uniform, that is, there exists $c_3 > 0$, such that*

$$\max_{l=1, \dots, d} \frac{\max(x_{l,j+1} - x_{l,j}, j = 0, \dots, N_l)}{\min(x_{l,j+1} - x_{l,j}, j = 0, \dots, N_l)} \leq c_3.$$

Furthermore, the number of interior knots $N_l \asymp n^{1/(2p+3)}$, where p is the degree of the spline and ' \asymp ' means both sides have the same order. In particular, $h \asymp n^{-1/(2p+3)}$.

- (A3) *For $1 \leq l \leq d$, the functions $\alpha_l \in C^p([0, 1])$, where $C^p([0, 1])$ denotes the space of p -times continuously differentiable functions on $[0, 1]$.*
- (A4) *The conditional variance function $\sigma^2(\mathbf{x}) = \text{Var}(Y|\mathbf{X} = \mathbf{x})$ is bounded on \mathcal{X} .*

S2. Auxiliary lemmas

Denote $\mathcal{M}_{n,0} \subset \mathcal{M}_n$ as

$$\mathcal{M}_{n,0} = \left\{ m_n(\mathbf{x}) = \sum_{l=1}^s g_l(x_l); g_l \in \varphi_l^{0,n} \right\},$$

which is the approximation space knowing $\alpha_{l0} = 0$, for $l = s + 1, \dots, d$. To prove Theorem 1, we will make use of two standard least square spline estimators, denoted as $\hat{m}_{n,0}^*$ and \hat{m}_n^* , which are the best least square approximation of m_0 in approximation spaces $\mathcal{M}_{n,0}$, and \mathcal{M}_n respectively. That is, define

$$\hat{m}_{n,0}^* = \operatorname{argmin}_{m_{n,0} \in \mathcal{M}_{n,0}} \|Y - m_{n,0}\|_n^2, \quad \hat{m}_n^* = \operatorname{argmin}_{m_n \in \mathcal{M}_n} \|Y - m_n\|_n^2.$$

Here we cite some results regarding the standard polynomial spline estimation.

Lemma 1 *Under conditions (A1-A4), with $\rho_n = 1/\sqrt{nh} + h^{p+1}$, one has*

$$(i) \quad \|\hat{m}_{n,0}^* - m_0\| = O_p(\rho_n), \quad \|\hat{m}_{n,0}^* - m_0\|_n = O_p(\rho_n).$$

$$(ii) \quad \|\hat{m}_n^* - m_0\| = O_p(\rho_n), \quad \|\hat{m}_n^* - m_0\|_n = O_p(\rho_n).$$

Lemma 1 is the standard results regarding the mean square (or L_2)-convergence rate for standard polynomial spline estimators (e.g. Theorem 1 in Huang 1998).

Lemma 2 *Under conditions (A1-A2), as $n \rightarrow \infty$, one has*

$$\sup_{\phi_1 \in \mathcal{M}_n, \phi_2 \in \mathcal{M}_n} \left| \frac{\langle \phi_1, \phi_2 \rangle_n - \langle \phi_1, \phi_2 \rangle}{\|\phi_1\| \|\phi_2\|} \right| = O_p \left(\sqrt{\frac{\log^2(n)}{nh}} \right).$$

In particular, there exist constants $0 < c < 1 < C$ such that, except on an event whose probability tends to zero as $n \rightarrow \infty$, $c\|m_n\| \leq \|m_n\|_n \leq C\|m_n\|, \forall m_n \in \mathcal{M}_n$.

Lemma 2 is crucial to prove both Theorem 1 and Theorem 2. It shows that the empirical and theoretical inner products are uniformly close over the approximation space \mathcal{M}_n . The general proof of Lemma 2 can be found in Xue and Yang (2006a) or Huang (1998).

Lemma 3 *Under condition (A1), let $\delta = (1 - c_1/c_2)^{1/2}$, and $c_4 = \left(\frac{1-\delta}{2}\right)^{(d-1)/2} > 0$. Then for any $m = \sum_{l=1}^d \alpha_l \in \mathcal{M}$, one has*

$$\|m\| \geq c_4 \sum_{l=1}^d \|\alpha_l\|.$$

Lemma 3 is the Lemma 1 in Stone (1985), which implies that the model space \mathcal{M} is essentially identifiable (up to sets of Lebesgue measure zero). That is, for any $m \in \mathcal{M}$, there is essentially a unique additive representation $m = \sum_{l=1}^d \alpha_l$, with $\alpha_l \in H_l^0$. The next lemma follows immediately from Lemmas 2 and 3.

Lemma 4 *Under conditions (A1-A2), there exists a constant $c_5 > 0$, such that, except on an event whose probability tends to zero, as $n \rightarrow \infty$, for any $m_n = \sum_{l=1}^d g_l \in \mathcal{M}_n$, with $g_l \in \varphi_l^{0,n}$,*

$$\|m_n\|_n \geq c_5 \sum_{l=1}^d \|g_l\|_n.$$