

KERNEL ADDITIVE SLICED INVERSE REGRESSION

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Supplementary Material

S1 Proof of the main theorem.

In the proofs, C denotes a generic positive constant. We first note that $E_{X^*}[(\hat{f}(X^*) - f(X^*))^2] = \|\Sigma^{1/2}(\hat{f} - f)\|_{\mathcal{H}}^2$. From (2.3), $\Sigma^{1/2}\hat{f}$ satisfies the eigenvalue equation

$$\Sigma^{1/2}(\Sigma_n + sI)^{-1}\Gamma_n\Sigma^{-1/2}(\Sigma^{1/2}\hat{f}) = \lambda\Sigma^{1/2}\hat{f}.$$

Similarly, $\Sigma^{1/2}f$ satisfies the eigenvalue equation

$$\Sigma^{1/2}\Sigma^{-1}\Gamma\Sigma^{-1/2}(\Sigma^{1/2}f) = \lambda\Sigma^{1/2}f.$$

Using the perturbation theory for operators, for example as in Chapter 4 of Kato (1995), we only need to show that $\|\Sigma^{1/2}(\Sigma_n + sI)^{-1}\Gamma_n\Sigma^{-1/2} - \Sigma^{1/2}\Sigma^{-1}\Gamma\Sigma^{-1/2}\|^2 = O_p(c_n n^{-d/(d+1)})$, where $\|A\|$ denotes the operator norm for an operator A defined on \mathcal{H}_K .

We write

$$\begin{aligned} & \|\Sigma^{1/2}(\Sigma_n + sI)^{-1}\Gamma_n\Sigma^{-1/2} - \Sigma^{1/2}\Sigma^{-1}\Gamma\Sigma^{-1/2}\| \\ &= \|\Sigma^{1/2}((\Sigma_n + sI)^{-1}\Gamma_n - (\Sigma + sI)^{-1}\Gamma + (\Sigma + sI)^{-1}\Gamma - \Sigma^{-1}\Gamma)\Sigma^{-1/2}\| \\ &= \|\Sigma^{1/2}((\Sigma + sI)^{-1}(\Gamma_n - \Gamma) + (\Sigma + sI)^{-1}(\Sigma - \Sigma_n)(\Sigma_n + sI)^{-1}\Gamma_n \\ & \quad + (\Sigma + sI)^{-1}\Gamma - \Sigma^{-1}\Gamma)\Sigma^{-1/2}\| \\ &\leq \|\Sigma^{1/2}(\Sigma + sI)^{-1}(\Gamma_n - \Gamma)\Sigma^{-1/2}\| + \|\Sigma^{1/2}(\Sigma + sI)^{-1}(\Sigma - \Sigma_n)(\Sigma_n + sI)^{-1}\Gamma_n\Sigma^{-1/2}\| \\ & \quad + \|\Sigma^{1/2}((\Sigma + sI)^{-1}\Gamma - \Sigma^{-1}\Gamma)\Sigma^{-1/2}\| \\ &=: (I) + (II) + (III). \end{aligned} \tag{S1.1}$$

To simplify the proofs and notations a little bit, we assume $\overline{K(\cdot, X)} = 0$ in the following, since using $\|\overline{K(\cdot, X)}\|_{\mathcal{H}} = O_p(n^{-1/2})$, such terms does not lead to extra difficulties in the proof. By the same reason, we also replace \hat{p}_h by $p_h = P(Y = y_h)$ in the following expression of Γ_n . Obviously we can rewrite Γ_n and Γ as

$$\begin{aligned} \Gamma_n &= \sum_{h=1}^H \frac{1}{p_h} \left(\frac{1}{n} \sum_{i=1}^n K(\cdot, X_i) I\{Y_i = y_h\} \right) \otimes \left(\frac{1}{n} \sum_{i=1}^n K(\cdot, X_i) I\{Y_i = y_h\} \right), \\ \Gamma &= \sum_{h=1}^H \frac{1}{p_h} E[(K(\cdot, X) I\{Y = y_h\})] \otimes E[K(\cdot, X) I\{Y = y_h\}]. \end{aligned} \tag{S1.2}$$

The term (III) is easy to deal with as follows. Using

$$\begin{aligned}
& \|\Sigma^{1/2}((\Sigma + sI)^{-1} - \Sigma^{-1})(E[K(\cdot, X)I\{Y = y_h\}] \otimes E[K(\cdot, X)I\{Y = y_h\}])\Sigma^{-1/2}\|^2 \\
&= \|s\Sigma^{1/2}(\Sigma + sI)^{-1}(\Sigma^{-1}E[K(\cdot, X)I\{Y = y_h\}]) \otimes (\Sigma^{-1/2}E[K(\cdot, X)I\{Y = y_h\}])\|^2 \\
&= O(\|s\Sigma^{1/2}(\Sigma + sI)^{-1}\|^2) \\
&= O(s),
\end{aligned}$$

we have $(III)^2 = O(s)$.

For the term (II), writing $\Sigma(x) = K(\cdot, x) \otimes K(\cdot, x)$ and using that

$$\begin{aligned}
& E\|\Sigma^{1/2}(\Sigma + sI)^{-1}\Sigma(x)\|_{HS}^2 \\
&= E\text{tr}(\Sigma(x)^2(\Sigma + sI)^{-1}\Sigma(\Sigma + sI)^{-1}) \\
&\leq CE\text{tr}(\Sigma(x)(\Sigma + sI)^{-1}\Sigma(\Sigma + sI)^{-1}) \\
&= C\text{tr}(\Sigma(\Sigma + sI)^{-1}\Sigma(\Sigma + sI)^{-1}) \\
&= C\sum_j \frac{\lambda_j^2}{(\lambda_j + s)^2},
\end{aligned}$$

where $\|\cdot\|_{HS}$ is the Hilbert-Schmidt norm. In the inequality above we used that $\|\Sigma(x)f\|_{\mathcal{H}} = \|K(\cdot, x)f(x)\|_{\mathcal{H}} = \sqrt{f^2(x)K(x, x)} \leq C\|f\|_{\infty} \leq C\|f\|_{\mathcal{H}}$ and thus $\|\Sigma(x)\| \leq C$, and the inequality $\text{tr}(AB) \leq \|A\|\text{tr}(B)$.

Thus using the Markov inequality, we have

$$\begin{aligned}
& (II)^2 \\
&\leq \|\Sigma^{1/2}(\Sigma + sI)^{-1}(\Sigma - \Sigma_n)\|^2 \cdot \|(\Sigma_n + sI)^{-1}\Gamma_n\Sigma^{-1/2}\|^2 \\
&= O_p\left(\sum_j \frac{\lambda_j^2}{n(\lambda_j + s)^2}\right) \|(\Sigma_n + sI)^{-1}\Gamma_n\Sigma^{-1/2}\|^2.
\end{aligned}$$

We will argue later that actually $\|(\Sigma_n + sI)^{-1}\Gamma_n\Sigma^{-1/2}\|^2 = O_p(1)$.

The term (I) is more complicated. Let Γ_{nh} and Γ_h be the terms on the right hand side of the sums in (S1.2) such that $\Gamma_n = \sum_h p_h^{-1}\Gamma_{nh}$, $\Gamma = \sum_h p_h^{-1}\Gamma_h$. To bridge Γ_{nh} and Γ_h , we further define

$$\Gamma'_{nh} = \left(\frac{1}{n}\sum_{i=1}^n K(\cdot, X_i)I\{Y_i = y_h\}\right) \otimes E[K(\cdot, X)I\{Y = y_h\}].$$

Since

$$\begin{aligned}
 & E\|\Sigma^{1/2}(\Sigma + sI)^{-1}(K(\cdot, X)I\{Y = y_h\} - E[K(\cdot, X)I\{Y = y_h\}]) \\
 & \quad \otimes (\Sigma^{-1/2}E[K(\cdot, X)I\{Y = y_h\}])\|_{HS}^2 \\
 \leq & CE\|\Sigma^{1/2}(\Sigma + sI)^{-1}(K(\cdot, X)I\{Y = y_h\} - E[K(\cdot, X)I\{Y = y_h\}])\|_{\mathcal{H}}^2 \\
 \leq & CE\|\Sigma^{1/2}(\Sigma + sI)^{-1}K(\cdot, X)I\{Y = y_h\}\|_{\mathcal{H}}^2 \\
 = & CE[I\{Y = y_j\}\langle \Sigma(\Sigma + sI)^{-2}K(\cdot, X), K(\cdot, X) \rangle_{\mathcal{H}}] \\
 = & CE[I\{Y = y_j\}\text{tr}(\Sigma(\Sigma + sI)^{-2}\Sigma(X))] \\
 = & CE[\text{tr}(\Sigma(\Sigma + sI)^{-2}\Sigma(X))E[I\{Y = y_j\}|X]] \\
 \leq & C\text{tr}(\Sigma^2(\Sigma + sI)^{-2}) \\
 = & C\sum_j \frac{\lambda_j^2}{(\lambda_j + s)^2},
 \end{aligned}$$

by Markov inequality,

$$\|\Sigma^{1/2}(\Sigma + sI)^{-1}(\Gamma'_{nh} - \Gamma_h)\Sigma^{-1/2}\|^2 = O_p\left(\sum_j \frac{\lambda_j^2}{n(\lambda_j + s)^2}\right).$$

Similarly one can show

$$\|\Sigma^{1/2}(\Sigma + sI)^{-1}(\Gamma_n - \Gamma'_{nh})\Sigma^{-1/2}\|^2 = O_p\left(\sum_j \frac{\lambda_j^2}{n(\lambda_j + s)^2}\right).$$

These imply that

$$(II)^2 = O_p\left(\sum_j \frac{\lambda_j^2}{n(\lambda_j + s)^2}\right).$$

Once we have shown $\|(\Sigma_n + sI)^{-1}\Gamma_n\Sigma^{-1/2}\| = O_p(1)$, the bounds given above will combine to obtain that (S1.1) is bounded by $O_p\left(s + \sum_j \frac{\lambda_j^2}{n(\lambda_j + s)^2}\right)$ and direct calculations by plugging in $\lambda_j \asymp j^{-d}$ and $s = c_n n^{-d/(d+1)}$ shows that this is $O_p(c_n n^{-d/(d+1)})$.

What is left is to show $\|(\Sigma_n + sI)^{-1}\Gamma_n\Sigma^{-1/2}\| = O_p(1)$, which is equivalent to showing $\|(\Sigma_n + sI)^{-1}\Gamma_n\Sigma^{-1/2} - \Sigma^{-1}\Gamma\Sigma^{-1/2}\| = O_p(1)$. Note this equation is actually similar to (S1.1). Following similar lines that are used to upper bound the terms (I)-(III) in (S1.1), we will get

$$\begin{aligned}
 & \|(\Sigma_n + sI)^{-1}\Gamma_n\Sigma^{-1/2} - \Sigma^{-1}\Gamma\Sigma^{-1/2}\|^2 \\
 = & O_p\left(\sum_j \frac{\lambda_j}{n(\lambda_j + s)^2}\right) \|(\Sigma_n + sI)^{-1}\Gamma_n\Sigma^{-1/2}\|^2 + O_p\left(1 + \sum_j \frac{\lambda_j}{n(\lambda_j + s)^2}\right).
 \end{aligned}$$

When $s = c_n n^{-d/(d+1)}$ with $c_n \rightarrow \infty$, by direct calculations we have $\sum_j \frac{\lambda_j}{n(\lambda_j + s)^2} = o(1)$ and thus the above displayed equation implies $\|(\Sigma_n + sI)^{-1}\Gamma_n\Sigma^{-1/2}\| = O_p(1)$. \square