

On Aggregate Dimension Reduction

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Supplementary Material

S1 Proof of Theorem 1

Let G be an open subset of $\Omega_{\mathbf{X}}$. Then, by part 1 of Proposition 11 in Yin et al. (2008), we have that $\mathcal{S}_{Y_G|\mathbf{X}_G} \subseteq \mathcal{S}_{Y|\mathbf{X}}$, which implies that $\text{span}\{\mathcal{S}_{Y_G|\mathbf{X}_G} : G \subseteq \Omega_{\mathbf{X}}\} \subseteq \mathcal{S}_{Y|\mathbf{X}}$.

By a result of Zhu and Zeng (2006), we have

$$\text{span}\{\partial h(y | \mathbf{x})/\partial \mathbf{x} : (\mathbf{x}, y) \in \Omega_{\mathbf{X}} \times \Omega_Y\} = \mathcal{S}_{Y|\mathbf{X}}. \quad (\text{S1.1})$$

Apply the same result to (\mathbf{X}_G, Y_G) to obtain

$$\text{span}\{\partial h(y | \mathbf{x})/\partial \mathbf{x} : (\mathbf{x}, y) \in G \times \Omega_Y\} = \mathcal{S}_{Y_G|\mathbf{X}_G}. \quad (\text{S1.2})$$

Now let (\mathbf{x}_0, y_0) be an arbitrary point in $(\Omega_{\mathbf{X}}, \Omega_Y)$, and let G be an open subset of $\Omega_{\mathbf{X}}$ that contains \mathbf{x}_0 . Then, by part 3 of Proposition 1, $h_G(y |$

$\mathbf{x}) = h(y | \mathbf{x})$ for all $(\mathbf{x}, y) \in G \times \Omega_Y$. Therefore, $[\partial h_G(y | \mathbf{x}) / \partial \mathbf{x}]_{\mathbf{x}_0, y_0} = [\partial h(y | \mathbf{x}) / \partial \mathbf{x}]_{\mathbf{x}_0, y_0}$. Thus, by (S1.1) and (S1.2) we have

$$\mathcal{S}_{Y|\mathbf{X}} \subseteq \cup\{\mathcal{S}_{Y|\mathbf{X}_G} : G \subseteq \Omega_{\mathbf{X}}\} \subseteq \text{span}\{\mathcal{S}_{Y|\mathbf{X}_G} : G \subseteq \Omega_{\mathbf{X}}\}. \quad (\text{S1.3})$$

Furthermore, by part 2 of Proposition 11 of Yin et al. (2008), there exists a compact set $K \subseteq \Omega_{\mathbf{X}}$ such that $\mathcal{S}_{Y_K|\mathbf{X}_K} = \mathcal{S}_{Y|\mathbf{X}}$, where (\mathbf{X}_K, Y_K) is defined as \mathbf{X} restricted on K . Since $\cup\{G : G \subseteq \Omega_{\mathbf{X}}\}$ forms an open cover of the compact set K , there is a finite subcover $\cup\{G_i : i = 1, \dots, m\}$ of K . Hence by the same argument leading to (S1.3) we have $\mathcal{S}_{Y|\mathbf{X}} \subseteq \cup\{\mathcal{S}_{Y|\mathbf{X}_{G_i}} : i = 1, \dots, m\}$, as desired. \square

S2 Proof of Theorem 2

We have

$$E(\mathbf{X}_G | Y_G = y) = \int_G \mathbf{x} \frac{h(y | \mathbf{x}) p_G(\mathbf{x})}{g_G(y)} d\mathbf{x} = \frac{1}{g_G(y)} \int_G \mathbf{x} h(y | \mathbf{x}) p_G(\mathbf{x}) d\mathbf{x}. \quad (\text{S2.1})$$

Let $\dot{h}(y | \mathbf{x})$ and $\ddot{h}(y | \mathbf{x})$ denote the first and second derivatives of h with respect to \mathbf{x} . By Taylor's theorem, for any $\mathbf{x} \in G$, there is a $\boldsymbol{\xi}$ with $\|\boldsymbol{\xi} - \boldsymbol{\mu}_G\| \leq \|G\|$ such that

$$h(y | \mathbf{x}) = h(y | \boldsymbol{\mu}_G) + \dot{h}^T(y | \boldsymbol{\mu}_G)(\mathbf{x} - \boldsymbol{\mu}_G) + \frac{1}{2}(\mathbf{x} - \boldsymbol{\mu}_G)^T \ddot{h}(y | \boldsymbol{\xi})(\mathbf{x} - \boldsymbol{\mu}_G). \quad (\text{S2.2})$$

In the meantime,

$$\begin{aligned}
 h(y \mid \boldsymbol{\mu}_G + \mathbf{P}_{\beta_G}(\mathbf{x} - \boldsymbol{\mu}_G)) &= h(y \mid \boldsymbol{\mu}_G) + \dot{h}^T(y \mid \boldsymbol{\mu}_G) \mathbf{P}_{\beta_G}(\mathbf{x} - \boldsymbol{\mu}_G) \\
 &\quad + \frac{1}{2}(\mathbf{x} - \boldsymbol{\mu}_G)^T \mathbf{P}_{\beta_G} \ddot{h}(y \mid \boldsymbol{\xi}) \mathbf{P}_{\beta_G}(\mathbf{x} - \boldsymbol{\mu}_G).
 \end{aligned} \tag{S2.3}$$

However, by construction, it is easy to see that $\dot{h}(y \mid \boldsymbol{\mu}_G) \in \text{span}(\mathbf{H}_G)$ almost everywhere in Ω_Y . Hence $\mathbf{P}_{\beta_G} \dot{h}(y \mid \boldsymbol{\mu}_G) = \dot{h}(y \mid \boldsymbol{\mu}_G)$. Because $\|\mathbf{x} - \boldsymbol{\mu}_G\| \leq \|G\|$ and the elements of $\ddot{h}(y \mid \boldsymbol{\xi})$ are bounded, the third terms on the right hand sides of (S2.2) and (S2.3) are of the order $O(\|G\|^2)$. Now subtract (S2.2) from (S2.3),

$$h(y \mid \mathbf{x}) = h(y \mid \boldsymbol{\mu}_G + \mathbf{P}_{\beta_G}(\mathbf{x} - \boldsymbol{\mu}_G)) + O(\|G\|^2) \text{ as } \|G\| \rightarrow 0. \tag{S2.4}$$

Substitute (S2.2) into the right hand side of (S2.1), using the relations $E(\mathbf{X}_G - \boldsymbol{\mu}_G) = 0$ and $\text{Var}(\mathbf{X}_G) = \boldsymbol{\Sigma}_G$, to obtain

$$\begin{aligned}
 E(\mathbf{X}_G - \boldsymbol{\mu}_G \mid Y_G = y) &= \frac{1}{g_G(y)} \boldsymbol{\Sigma}_G \dot{h}(y \mid \boldsymbol{\mu}_G) \\
 &\quad + \frac{1}{2g_G(y)} \int_G [(\mathbf{x} - \boldsymbol{\mu}_G)(\mathbf{x} - \boldsymbol{\mu}_G)^T \ddot{h}(y \mid \boldsymbol{\xi})(\mathbf{x} - \boldsymbol{\mu}_G)] p_G(\mathbf{x}) d\mathbf{x}.
 \end{aligned} \tag{S2.5}$$

Since $\|\mathbf{x} - \boldsymbol{\mu}_G\| \leq \|G\|$ and the components of $\ddot{h}(y \mid \boldsymbol{\xi})$ are bounded, the second term on the right is of the order $O(\|G\|^3)$. In other words,

$$E(\mathbf{X}_G - \boldsymbol{\mu}_G \mid Y_G = y) = \frac{1}{g_G(y)} \boldsymbol{\Sigma}_G \dot{h}(y \mid \boldsymbol{\mu}_G) + O(\|G\|^3).$$

Multiply both sides by Σ_G^{-1} , keeping in mind that $\Sigma_G = O(\|G\|^2)$, to obtain

$$\Sigma_G^{-1}E(\mathbf{X}_G - \boldsymbol{\mu}_G \mid Y_G = y) = \frac{1}{g_G(y)}\dot{h}(y \mid \boldsymbol{\mu}_G) + O(\|G\|). \quad (\text{S2.6})$$

Meanwhile, if we multiply both sides of the above equality by \mathbf{P}_{β_G} , then, because $\dot{h}(y \mid \boldsymbol{\mu}_G) \in \text{span}(\beta_G)$ for almost every $y \in \Omega_Y$, we have

$$\mathbf{P}_{\beta_G}\Sigma_G^{-1}E(\mathbf{X}_G - \boldsymbol{\mu}_G \mid y) = \frac{1}{g_G(y)}\dot{h}(y \mid \boldsymbol{\mu}_G) + O(\|G\|). \quad (\text{S2.7})$$

Now subtract (S2.7) from (S2.6) to prove (3.1). \square

S3 Proof of Theorem 3

Let $\boldsymbol{\mu}_G^*$ be the center of G . Since p_G has bounded derivative, $p_G(\mathbf{x}) = p_G(\boldsymbol{\mu}_G^*) + O(\|G\|)$. Hence

$$\begin{aligned} \boldsymbol{\mu}_G &= \int_G (\mathbf{x} - \boldsymbol{\mu}_G^* + \boldsymbol{\mu}_G^*)p_G(\mathbf{x})d\mathbf{x} \\ &= \boldsymbol{\mu}_G^* + \int_G (\mathbf{x} - \boldsymbol{\mu}_G^*)[p_G(\boldsymbol{\mu}_G^*) + O(\|G\|)]d\mathbf{x} = \boldsymbol{\mu}_G^* + O(\|G\|^3). \end{aligned}$$

Hence the integral in the second term on the right hand side of (S2.5) is

$$\begin{aligned} &\int_G [(\mathbf{x} - \boldsymbol{\mu}_G^*)(\mathbf{x} - \boldsymbol{\mu}_G^*)^T \ddot{h}(y \mid \boldsymbol{\xi})(\mathbf{x} - \boldsymbol{\mu}_G^*) + O(\|G\|^5)][p_G(\boldsymbol{\mu}_G^*) + O(\|G\|)]d\mathbf{x} \\ &= \int_G [(\mathbf{x} - \boldsymbol{\mu}_G^*)(\mathbf{x} - \boldsymbol{\mu}_G^*)^T \ddot{h}(y \mid \boldsymbol{\xi})(\mathbf{x} - \boldsymbol{\mu}_G^*)]p_G(\boldsymbol{\mu}_G^*)d\mathbf{x} + O(\|G\|^5) \end{aligned}$$

However, the leading term on the right is also of the order $O(\|G\|^5)$, because

$$\begin{aligned} & \int_G [(\mathbf{x} - \boldsymbol{\mu}_G^*)(\mathbf{x} - \boldsymbol{\mu}_G^*)^T \ddot{h}(y \mid \boldsymbol{\xi})(\mathbf{x} - \boldsymbol{\mu}_G^*)] p_G(\boldsymbol{\mu}_G^*) d\mathbf{x} \\ &= p_G(\boldsymbol{\mu}_G^*) \int_G (\mathbf{x} - \boldsymbol{\mu}_G^*)(\mathbf{x} - \boldsymbol{\mu}_G^*)^T \ddot{h}(y \mid \boldsymbol{\mu}_G)(\mathbf{x} - \boldsymbol{\mu}_G^*) d\mathbf{x} + O(\|G\|^5), \end{aligned}$$

where the first term is 0 since G is an open ball. The rest of the proof is to the argument following (S2.5). \square

Bibliography

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