

# High-dimensional Multivariate Mediation with Application to Neuroimaging Data

## Supplementary Materials

### Proofs

**Lemma 1 (Consistency Theorem):** Suppose that  $Q(\boldsymbol{\theta}; \mathbf{Z}_n)$  is continuous in  $\boldsymbol{\theta}$  and there exists a function  $Q_0(\boldsymbol{\theta})$  such that:  $Q_0(\boldsymbol{\theta})$  is uniquely maximized at  $\boldsymbol{\theta}_0$ ;  $\Theta$  is compact;  $Q_0(\boldsymbol{\theta})$  is continuous in  $\boldsymbol{\theta}$ ; and  $Q(\boldsymbol{\theta}; \mathbf{Z}_n)$  converges uniformly in probability to  $Q_0(\boldsymbol{\theta})$ . Then  $\hat{\boldsymbol{\theta}}(\mathbf{Z}_n)$  defined as the value of  $\boldsymbol{\theta} \in \Theta$  which for each  $\mathbf{Z}_n = \mathbf{z}_n$  maximizes the objective function  $Q(\boldsymbol{\theta}; \mathbf{Z}_n)$  satisfies  $\hat{\boldsymbol{\theta}}(\mathbf{Z}_n) \xrightarrow{p} \boldsymbol{\theta}_0$ .

**Proof of Lemma 1:** See Theorem 2.1 in [\(Newey and McFadden, 1994\)](#). ■

**Lemma 2:** Consider a compact space  $\Theta$ . Let

$$L(z, \mathbf{w}, \boldsymbol{\theta}, \lambda) = f^{\text{obj}}(z, \mathbf{w}, \boldsymbol{\theta}) + f^{\text{pen}}(z, \mathbf{w}, \lambda),$$

where  $f^{\text{obj}}(z, \mathbf{w}, \boldsymbol{\theta})$  is an objective function and  $f^{\text{pen}}(z, \mathbf{w}, \lambda) = \frac{\lambda\{f^{\text{cons}}(\mathbf{w}) - c\}}{n}$  is a penalization function, for some constant  $c$ . If both the objective function and the penalization function can be profiled by  $\boldsymbol{\theta}$ , defined as  $f^{\text{obj}}(z, \boldsymbol{\theta})$  and  $f^{\text{pen}}(\boldsymbol{\theta}) := \frac{\lambda(\boldsymbol{\theta})\{f^{\text{cons}}(\boldsymbol{\theta}) - c\}}{n}$ ; the objective function is a log likelihood function; both  $f^{\text{obj}}(z, \boldsymbol{\theta})$  and  $f^{\text{pen}}(\boldsymbol{\theta})$  are continuous in  $\boldsymbol{\theta}$ ; and there exists a function  $d_0(z)$  such that  $|L(z, \boldsymbol{\theta})| := |f^{\text{obj}}(z, \boldsymbol{\theta}) + f^{\text{pen}}(\boldsymbol{\theta})| \leq d_0(z)$  for all  $\boldsymbol{\theta} \in \Theta$  and  $z \in \mathcal{Z}$ , and  $\mathbb{E}_{\boldsymbol{\theta}_0}[d_0(x)] < \infty$ , then

- i.  $q_0(\boldsymbol{\theta}) := \mathbb{E}_{\boldsymbol{\theta}_0}[L(z, \boldsymbol{\theta})]$  is continuous in  $\boldsymbol{\theta}$ ;
- ii.  $\sup_{\boldsymbol{\theta} \in \Theta} |q(\boldsymbol{\theta}; \mathbf{Z}_n) - q_0(\boldsymbol{\theta})| \xrightarrow{P} 0$ , where  $q(\boldsymbol{\theta}; \mathbf{Z}_n) := \frac{1}{n} L(\mathbf{Z}_n, \boldsymbol{\theta})$ .

Note: the above Lemma can be stated in a more general case where there are multiple sets of parameters and several constraint functions.

**Proof of Lemma 2:** Consider the regularity conditions stated in the Appendix.

$\forall \boldsymbol{\theta} \in \Theta$ , choose a sequence  $\boldsymbol{\theta}_k \in \Theta$ , such that  $\boldsymbol{\theta}_k \rightarrow \boldsymbol{\theta}$ . By (N-3), we have  $L(x; \boldsymbol{\theta}_k) \rightarrow L(x; \boldsymbol{\theta})$ . By (N-4) and the dominated convergence theorem (DCT),  $q_0(\boldsymbol{\theta}_k) := \mathbb{E}_{\boldsymbol{\theta}_0}(L(Z, \boldsymbol{\theta}_k)) \rightarrow \mathbb{E}_{\boldsymbol{\theta}_0}(L(Z, \boldsymbol{\theta})) = q_0(\boldsymbol{\theta})$ . Hence,  $q_0(\boldsymbol{\theta})$  is continuous in  $\boldsymbol{\theta}$ .

$L(z, \boldsymbol{\theta})$  is uniformly continuous since  $f^{\text{obj}}(z, \boldsymbol{\theta})$  and  $f^{\text{prof}}(\boldsymbol{\theta})$  are continuous in  $\boldsymbol{\theta}$ . Hence,

$$\Delta(z, \delta) = \sup_{\{(\boldsymbol{\theta}_1, \boldsymbol{\theta}_2) : \|\boldsymbol{\theta}_1 - \boldsymbol{\theta}_2\| < \delta\}} |L(z, \boldsymbol{\theta}_1) - L(z, \boldsymbol{\theta}_2)| \rightarrow 0$$

as  $\delta \rightarrow 0$ . By (N-4),  $\Delta(z, \delta) \leq 2d_0(z)$ ,  $\forall \delta$ . By DCT,  $\mathbb{E}_{\boldsymbol{\theta}_0}[\Delta(Z, \delta)] \rightarrow 0$  as  $\delta \rightarrow 0$ .

Define  $B(\boldsymbol{\theta}_j, \delta) = \{\tilde{\boldsymbol{\theta}} : \|\tilde{\boldsymbol{\theta}} - \boldsymbol{\theta}_j\| < \delta\}$ . Since  $\Theta$  is compact, for every fixed  $\delta$ ,  $\exists$  a subcover  $\{B(\boldsymbol{\theta}_j, \delta), j = 1, \dots, J\}$  such that  $\bigcup_{j=1}^{J < \infty} B(\boldsymbol{\theta}_j, \delta) \supset \Theta$ . Then, we have:

$$|q(\boldsymbol{\theta}; \mathbf{Z}_n) - q_0(\boldsymbol{\theta})| \leq |q(\boldsymbol{\theta}; \mathbf{Z}_n) - q(\boldsymbol{\theta}_j; \mathbf{Z}_n)| \quad (1)$$

$$+ |q(\boldsymbol{\theta}_j; \mathbf{Z}_n) - q_0(\boldsymbol{\theta}_j)| \quad (2)$$

$$+ |q_0(\boldsymbol{\theta}_j) - q_0(\boldsymbol{\theta})|. \quad (3)$$

Choose  $\boldsymbol{\theta}_j$  such that  $\boldsymbol{\theta} \in B(\boldsymbol{\theta}_j; \delta)$ . Since  $\|\boldsymbol{\theta} - \boldsymbol{\theta}_j\| < \delta$ , then:

$$(1) = \left| \frac{1}{n} \sum_{i=1}^n \{L(Z_i, \boldsymbol{\theta}) - L(Z_i, \boldsymbol{\theta}_j)\} \right| \leq \frac{1}{n} \sum_{i=1}^n |L(Z_i, \boldsymbol{\theta}) - L(Z_i, \boldsymbol{\theta}_j)| \leq \frac{1}{n} \sum_{i=1}^n \Delta(Z_i, \delta).$$

Next, (2)  $< \max_{j \in \{1, \dots, J\}} |q(\boldsymbol{\theta}_j; \mathbf{Z}_n) - q_0(\boldsymbol{\theta}_j)|$ ; and choose  $\delta$  to be small, then (3)  $\leq \sup_{\{(\boldsymbol{\theta}_1, \boldsymbol{\theta}_2) : \|\boldsymbol{\theta}_1 - \boldsymbol{\theta}_2\| < \delta\}} |q_0(\boldsymbol{\theta}_1) - q_0(\boldsymbol{\theta}_2)| \leq \epsilon^*(\delta)$ , where  $\epsilon(\delta) \rightarrow 0$  as  $\delta \rightarrow 0$ .

Combining (1) - (3), we have:

$$\sup_{\boldsymbol{\theta} \in \Theta} |q(\boldsymbol{\theta}; \mathbf{Z}_n) - q_0(\boldsymbol{\theta})| \leq \frac{1}{n} \sum_{i=1}^n \Delta(Z_i, \delta) + \max_{j \in \{1, \dots, J\}} |q(\boldsymbol{\theta}_j; \mathbf{Z}_n) - q_0(\boldsymbol{\theta}_j)| + \epsilon^*(\delta).$$

Choose  $\delta_1 \in \{\delta : \epsilon^*(\delta) \leq \frac{\epsilon}{3}\}$ . Then for any  $\delta < \delta_1$ , we have:

$$\begin{aligned} P_{\boldsymbol{\theta}_0}[\sup_{\boldsymbol{\theta} \in \Theta} |q(\boldsymbol{\theta}; \mathbf{Z}_n) - q_0(\boldsymbol{\theta})| > \epsilon] \\ \leq P_{\boldsymbol{\theta}_0}[\frac{1}{n} \sum_{i=1}^n \Delta(Z_i, \delta) + \max_{j \in \{1, \dots, J\}} |q(\boldsymbol{\theta}_j; \mathbf{Z}_n) - q_0(\boldsymbol{\theta}_j)| > \frac{2\epsilon}{3}] \\ \leq P_{\boldsymbol{\theta}_0}[\frac{1}{n} \sum_{i=1}^n \Delta(Z_i, \delta) > \frac{\epsilon}{3}] + \end{aligned} \quad (4a)$$

$$P_{\boldsymbol{\theta}_0}[\max_{j \in \{1, \dots, J\}} |q(\boldsymbol{\theta}_j; \mathbf{Z}_n) - q_0(\boldsymbol{\theta}_j)| > \frac{\epsilon}{3}] \quad (4b)$$

Note that (4a) =  $P_{\boldsymbol{\theta}_0}[\frac{1}{n} \sum_{i=1}^n \{\Delta(Z_i, \delta) - \mathbb{E}_{\boldsymbol{\theta}_0}[\Delta(Z; \delta)]\} + \mathbb{E}_{\boldsymbol{\theta}_0}[\Delta(Z; \delta)] > \frac{\epsilon}{3}]$ , where  $\mathbb{E}_{\boldsymbol{\theta}_0}[\Delta(Z; \delta)] \rightarrow 0$  as  $\delta \rightarrow 0$ . Choose  $\delta_2 \in \{\delta : \mathbb{E}_{\boldsymbol{\theta}_0}[\Delta(Z; \delta)] < \frac{\epsilon}{6}\}$ . Take  $\delta < \min(\delta_1, \delta_2)$ . Then:

$$P_{\boldsymbol{\theta}_0}[\frac{1}{n} \sum_{i=1}^n \{\Delta(Z_i, \delta) - \mathbb{E}_{\boldsymbol{\theta}_0}[\Delta(Z; \delta)]\} > \frac{\epsilon}{6}] := (4)'$$

By the Weak Law of Large Numbers (WLLN),  $\exists N_1(\epsilon, \xi)$  such that  $\forall n > N_1(\epsilon, \xi)$ ,  $(4)' < \frac{\xi}{2}$ .

Consider the finite subcover  $\{B(\boldsymbol{\theta}_j, \delta), j = 1, \dots, J\}$  for  $\delta$  considered above. Note that:

$$(5) = P_{\boldsymbol{\theta}_0}[\bigcup_{j=1}^J \{|q(\boldsymbol{\theta}_j; \mathbf{Z}_n) - q_0(\boldsymbol{\theta}_j)| > \frac{\epsilon}{3}\}] \leq \sum_{j=1}^J P_{\boldsymbol{\theta}_0}[|q(\boldsymbol{\theta}_j; \mathbf{Z}_n) - q_0(\boldsymbol{\theta}_j)| > \frac{\epsilon}{3}].$$

By the WLLN,  $\forall \boldsymbol{\theta}_j$  and  $\forall \epsilon, \xi > 0$ ,  $\exists N_{2j}(\epsilon, \xi)$  such that  $\forall n > N_{2j}(\epsilon, \xi)$ :  $P_{\boldsymbol{\theta}_0}[|q(\boldsymbol{\theta}_j; \mathbf{Z}_n) - q_0(\boldsymbol{\theta}_j)| > \frac{\epsilon}{3}] \leq \frac{\xi}{2J}$ . Let  $N_2(\epsilon, \xi) = \max_{j \in \{1, \dots, J\}} \{N_{2j}\}$ . Then,  $\forall n > N_2(\epsilon, \xi)$ , we have:  $\sum_{j=1}^J P_{\boldsymbol{\theta}_0}[|q(\boldsymbol{\theta}_j; \mathbf{Z}_n) - q_0(\boldsymbol{\theta}_j)| > \frac{\epsilon}{3}] < \frac{\xi}{2}$ . Hence, (4b)  $< \frac{\xi}{2}$ .

(4a) and (4b) show that  $\exists$  an  $N(\epsilon, \xi) = \max(N_1(\epsilon, \xi), N_2(\epsilon, \xi))$  such that  $\forall n > N(\epsilon, \xi)$ ,

$$P_{\boldsymbol{\theta}_0}[\sup_{\boldsymbol{\theta} \in \Theta} |q(\boldsymbol{\theta}; \mathbf{Z}_n) - q_0(\boldsymbol{\theta})| < \xi]. \quad \blacksquare$$

### Proof of Theorem 1:

Define  $Q(\boldsymbol{\theta}; \mathbf{Z}_n) := \frac{1}{n} L(\boldsymbol{\theta}; \mathbf{Z}_n)$  and  $Q_0(\boldsymbol{\theta}) := \mathbb{E}_{\boldsymbol{\theta}_0}(L(\boldsymbol{\theta}_0; \mathbf{Z}_n))$ ,  $L(\boldsymbol{\theta}; z) := f^{\text{obj}}(\boldsymbol{\theta}; z) + \frac{\lambda(\boldsymbol{\theta})(f^{\text{cons}} - c)}{n}$ , and  $\hat{\boldsymbol{\theta}}(\mathbf{Z}_n) := \operatorname{argmax}_{\boldsymbol{\theta}} \{L(\mathbf{Z}_n; \boldsymbol{\theta})\}$ , henceforth  $\hat{\boldsymbol{\theta}}$ .

## I. Consistency

To show

$$\hat{\boldsymbol{\theta}} \xrightarrow{p} \boldsymbol{\theta}_0, \quad (5)$$

by Lemma 1, it suffices to show: (a)  $Q(\boldsymbol{\theta}; \mathbf{Z}_n)$  is continuous in  $\boldsymbol{\theta}$ ; (b)  $Q_0(\boldsymbol{\theta})$  is continuous in  $\boldsymbol{\theta}$ ; and (c)  $\sup_{\boldsymbol{\theta} \in H} \|Q(\boldsymbol{\theta}; \mathbf{Z}_n) - Q_0(\boldsymbol{\theta})\| \xrightarrow{p} 0$ . Note that (a) is implied by (N-3); (N-3), (N-4), and Lemma 2 give (d)  $q_0(\boldsymbol{\theta}) := \mathbb{E}_{\boldsymbol{\theta}_0}(L(z, \boldsymbol{\theta}))$  is continuous in  $\boldsymbol{\theta}$ ; and (e)  $\sup_{\boldsymbol{\theta} \in H} \|\hat{q}(\boldsymbol{\theta}; \mathbf{Z}_n) - q_0(\boldsymbol{\theta})\| \xrightarrow{p} 0$ , where  $\hat{q}(\boldsymbol{\theta}; \mathbf{Z}_n) = \frac{1}{n} \sum_{i=1}^n L(Z_i, \boldsymbol{\theta})$ . Then (d) implies (b); (e) implies (c).  $\blacksquare$

## II. Asymptotic Normality

Let  $\ell(z; \boldsymbol{\theta}) := \frac{\partial f^{\text{obj}}(z; \boldsymbol{\theta})}{\partial \boldsymbol{\theta}} + \frac{\nabla^\theta \lambda(\boldsymbol{\theta})(f^{\text{cons}}(\boldsymbol{\theta}) - c) + \lambda(\boldsymbol{\theta}) \nabla^\theta f^{\text{cons}}(\boldsymbol{\theta})}{n}$ ,  $\hat{q}(\mathbf{Z}_n; \boldsymbol{\theta}) = \frac{1}{n} \sum_{i=1}^n L(Z_i; \boldsymbol{\theta})$ ,  $q_0(\boldsymbol{\theta}) = \mathbb{E}_{\boldsymbol{\theta}_0}(L(z; \boldsymbol{\theta}))$ ,  $\hat{q}(\mathbf{Z}_n; \boldsymbol{\theta}) = \frac{1}{n} \sum_{i=1}^n \ell(Z_i; \boldsymbol{\theta})$ , and  $\dot{q}_0(\boldsymbol{\theta}) = \mathbb{E}_{\boldsymbol{\theta}_0}(\ell(z; \boldsymbol{\theta}))$ .

$\hat{\boldsymbol{\theta}}$  satisfies:  $\frac{\partial L(\mathbf{Z}_n, \boldsymbol{\theta})}{\partial \boldsymbol{\theta}} /_{\boldsymbol{\theta}=\hat{\boldsymbol{\theta}}} = 0$ . Hence,  $\hat{q}(\mathbf{Z}_n; \hat{\boldsymbol{\theta}}) = 0$ . (N-2) and (N-6) give,

$$\dot{q}_0(\boldsymbol{\theta}) = o_p(n^{-1/2}). \quad (6)$$

Expanding  $\hat{q}(\mathbf{Z}_n; \hat{\boldsymbol{\theta}})$ , we have:

$$0 = \hat{q}(\mathbf{Z}_n; \hat{\boldsymbol{\theta}}) = \hat{q}(\mathbf{Z}_n; \boldsymbol{\theta}_0) + D_n^*(\mathbf{Z}_n)(\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}_0) \quad (7)$$

where  $D_n^*(\mathbf{Z}_n) = \frac{\partial \hat{q}(\mathbf{Z}_n; \boldsymbol{\theta})}{\partial \boldsymbol{\theta}} /_{\boldsymbol{\theta}=\boldsymbol{\theta}^*}$ , for some  $\boldsymbol{\theta}^*$  in the open interval bounded by  $\boldsymbol{\theta}_0$  and  $\hat{\boldsymbol{\theta}}$ .

(N-9), (N-10) and Lemma 2 give:  $\sup \|\hat{D}(\mathbf{Z}_n; \boldsymbol{\theta}) - D_0(\boldsymbol{\theta})\| \xrightarrow{p} 0$ , where  $D_0(\boldsymbol{\theta}) = \mathbb{E}_{\boldsymbol{\theta}_0}(\nabla^\theta \ell(z; \boldsymbol{\theta}))$  and  $D_0(\boldsymbol{\theta}_0) = \mathbb{E}_{\boldsymbol{\theta}_0}(\nabla^\theta \ell(z; \boldsymbol{\theta}_0))$ .

Since  $\boldsymbol{\theta}^* \in \mathcal{N}_r(\boldsymbol{\theta}_0)$  w.p.1., then  $D_n^*(\mathbf{Z}_n) \xrightarrow{p} D_0(\boldsymbol{\theta}_0)$ . From (7), we have  $\sqrt{n}(\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}_0) = -\{D_n^*(\mathbf{Z}_n)\}^{-1} \{\sqrt{n}\hat{q}(\mathbf{Z}_n; \boldsymbol{\theta}_0)\}$ ; From (6), we have  $\sqrt{n}\hat{q}(\mathbf{Z}_n; \boldsymbol{\theta}_0) = \sqrt{n} \left( \frac{\sum_{i=1}^n \ell(Z_i; \boldsymbol{\theta}_0)}{n} - \right.$

$\dot{q}_0(\boldsymbol{\theta})\Big) + \sqrt{n}\dot{q}_0(\boldsymbol{\theta}) \rightarrow N(0, V(\boldsymbol{\theta}_0))$ , where  $V(\boldsymbol{\theta}_0) = \mathbb{E}_{\boldsymbol{\theta}_0}\left(\ell(z; \boldsymbol{\theta})\ell^\top(z; \boldsymbol{\theta})\right)$ . It follows:

$$\sqrt{n}(\hat{\boldsymbol{\theta}} - \boldsymbol{\theta}_0) \rightarrow N(0, \Sigma(\boldsymbol{\theta}_0))$$

where  $\Sigma(\boldsymbol{\theta}_0) = D_0^{-1}(\boldsymbol{\theta}_0)V(\boldsymbol{\theta}_0)[D_0^{-1}(\boldsymbol{\theta}_0)]^\top$ . ■

## Proof of Theorem 2:

### I. Consistency

(N-0), and Eqs. (5)-(7) in Section 3.3 show that, for every fixed estimate  $\hat{\boldsymbol{\theta}}$ , there is a unique  $\mathbf{w} : \mathbb{R}^{\dim(\hat{\boldsymbol{\theta}})} \mapsto \mathbb{R}^{\dim(\mathbf{w})}$ , such that  $\mathbf{w} = \mathbf{w}(\hat{\boldsymbol{\theta}})$ . Hence, under (N-13), (5) gives:

$$\mathbf{w}(\hat{\boldsymbol{\theta}}) \xrightarrow{p} \mathbf{w}(\boldsymbol{\theta}_0).$$

### II. Asymptotic Normality

Under (N-12) and (N-13), and by the Multivariate Delta Method,

$$\sqrt{n}(\mathbf{w}(\hat{\boldsymbol{\theta}}) - \mathbf{w}(\boldsymbol{\theta}_0)) \rightarrow N\left(0, [\nabla \mathbf{w}(\boldsymbol{\theta})]^\top \Sigma(\boldsymbol{\theta}_0) \nabla \mathbf{w}(\boldsymbol{\theta})\right). \quad \blacksquare$$

## References and Notes

Newey, W. K. and D. McFadden (1994). Large sample estimation and hypothesis testing. *Handbook of econometrics* 4, 2111–2245.