

## Error-Correction Factor Models for High-dimensional Cointegrated Time Series

Yundong Tu, Qiwei Yao and Rongmao Zhang

*Peking University, London School of Economics and Zhejiang University*

### Supplementary Material

#### Lemmas and Technical Proofs

**Lemma 5.** *Under Condition 1 or conditions of Theorem 3, we have*

$$\frac{1}{n} \sum_{t=1}^{n-1} (\widehat{\mathbf{A}}_2' \mathbf{y}_t \mathbf{y}_t' \widehat{\mathbf{A}}_2 - \mathbf{A}_2' \mathbf{y}_t \mathbf{y}_t' \mathbf{A}_2) = o_p(1). \quad (\text{S0.1})$$

*Proof.* We first show the case with fixed  $p$ . Since  $\{\mathbf{x}_{t2}, \mathbf{f}_t, \boldsymbol{\varepsilon}_t\}$  is  $\alpha$  mixing with mixing coefficients  $\alpha_m$  satisfying

$$\sum_{m=1}^{\infty} \alpha_m^{1-1/\gamma} < \infty, \quad (\text{S0.2})$$

it follows that  $\{\nabla \mathbf{y}_t\}$  is a  $\alpha$  mixing process with mixing coefficients satisfying (S0.2). Thus, by Theorem 3.2.3 of Lin and Lu (1997), there exists a  $p$ -dimensional Gaussian process  $\mathbf{g}(t)$  such that

$$\mathbf{y}_{[nt]}/\sqrt{n} \xrightarrow{d} \mathbf{g}(t), \text{ on } D[0, 1]. \quad (\text{S0.3})$$

From (S0.3) and the continuous mapping theorem, it follows that

$$\frac{1}{n^2} \sum_{t=1}^n \mathbf{y}_t \mathbf{y}'_t \xrightarrow{d} \int_0^1 \mathbf{g}(t) \mathbf{g}'(t) dt. \quad (\text{S0.4})$$

Further, by  $E\|\mathbf{x}_{t2}\|^{2\gamma} < \infty$  for some  $\gamma > 1$ , we have

$$\max_{1 \leq t \leq n} \|\mathbf{x}_{t2} - E\mathbf{x}_{t2}\|/\sqrt{n} = o_p(1), \text{ and } \frac{1}{n} \sum_{t=1}^n \|\mathbf{x}_{t2} - E\mathbf{x}_{t2}\| = O_p(1). \quad (\text{S0.5})$$

Combining (S0.3) and (S0.5) (see Lemma 7 of ZRY) yields

$$\frac{1}{n^{3/2}} \left\| \sum_{t=1}^n \mathbf{y}_t \mathbf{x}'_{t2} \right\|_2 = o_p(1). \quad (\text{S0.6})$$

On the other hand, by  $\nabla_{\mathbf{x}_{t1}} = \mathbf{A}'_1 \nabla_{\mathbf{y}_t}$ , we know  $(\nabla_{\mathbf{x}_{t1}}, \mathbf{x}_{t2})$  is also  $\alpha$  mixing with mixing coefficients satisfying (S0.2). As a result, by the proof of Theorem 1 in ZRY,

$$\|\widehat{\mathbf{A}}_2 - \mathbf{A}_2\|_2 = O_p(1/n). \quad (\text{S0.7})$$

By (S0.4), (S0.6) and (S0.7), we have

$$\begin{aligned} & \left\| \frac{1}{n} \sum_{t=1}^{n-1} (\widehat{\mathbf{A}}'_2 \mathbf{y}_t \mathbf{y}'_t \widehat{\mathbf{A}}_2 - \mathbf{A}'_2 \mathbf{y}_t \mathbf{y}'_t \mathbf{A}_2) \right\|_2 \\ &= \left\| (\widehat{\mathbf{A}}_2 - \mathbf{A}_2)' \frac{\sum_{t=1}^{n-1} \mathbf{y}_t (\mathbf{A}'_2 \mathbf{y}_t)'}{n} + \frac{\sum_{t=1}^{n-1} (\mathbf{A}'_2 \mathbf{y}_t) \mathbf{y}'_t}{n} (\widehat{\mathbf{A}}_2 - \mathbf{A}_2) \right. \\ & \quad \left. + (\widehat{\mathbf{A}}_2 - \mathbf{A}_2)' \frac{\sum_{t=1}^{n-1} \mathbf{y}_t \mathbf{y}'_t}{n} (\widehat{\mathbf{A}}_2 - \mathbf{A}_2) \right\|_2 \\ &= \left\| (\widehat{\mathbf{A}}_2 - \mathbf{A}_2)' \frac{\sum_{t=1}^{n-1} \mathbf{y}_t \mathbf{x}'_{t2}}{n} + \frac{\sum_{t=1}^{n-1} \mathbf{x}_{t2} \mathbf{y}'_t}{n} (\widehat{\mathbf{A}}_2 - \mathbf{A}_2) + (\widehat{\mathbf{A}}_2 - \mathbf{A}_2) \frac{\sum_{t=1}^{n-1} \mathbf{y}_t \mathbf{y}'_t}{n} (\widehat{\mathbf{A}}_2 - \mathbf{A}_2)' \right\|_2 \\ &= o_p(1). \end{aligned} \quad (\text{S0.8})$$

Next, consider the case  $p = o(n^c)$ . Let  $\boldsymbol{\varsigma}_t$  be a  $k$ -dimensional  $I(1)$  process such that  $\nabla \boldsymbol{\varsigma}_t = \mathbf{v}_t$ . By Remark 2 of ZRY, we know that Condition 3 (i) and Remark 3 of ZRY hold for  $\boldsymbol{\varsigma}_t$ . Let  $\mathbf{M}_1, \mathbf{M}_2$  be  $k \times (p-r)$  and  $k \times r$

matrices such that  $\mathbf{M}$  given in (i) of Condition 3 satisfying  $\mathbf{M}' = (\mathbf{M}_1, \mathbf{M}_2)$ . Let  $\mathbf{F}(t) = (F^1(t), \dots, F^k(t))'$  be defined as in ZRY and  $\bar{\boldsymbol{\varsigma}} = \frac{1}{n} \sum_{t=1}^n \boldsymbol{\varsigma}_t$ , then

$$\begin{aligned} & \left\| \frac{1}{n^2} \sum_{t=1}^n (\mathbf{x}_{t1} - \bar{\mathbf{x}}_1)(\mathbf{x}_{t1} - \bar{\mathbf{x}}_1)' - \mathbf{M}'_1 \int_0^1 \mathbf{F}(t)\mathbf{F}'(t) dt \mathbf{M}_1 \right\|_2 \\ &= \left\| \mathbf{M}'_1 \left( \frac{1}{n^2} \sum_{t=1}^n (\boldsymbol{\varsigma}_t - \bar{\boldsymbol{\varsigma}})(\boldsymbol{\varsigma}_t - \bar{\boldsymbol{\varsigma}})' - \int_0^1 \mathbf{F}(t)\mathbf{F}'(t) dt \right) \mathbf{M}_1 \right\|_2 = o_p(1). \end{aligned} \quad (\text{S0.9})$$

By Remark 3 of ZRY, we have  $\lambda_{\min} \left( \int_0^1 \mathbf{F}(t)\mathbf{F}'(t) dt \right) \geq 1/k$  in probability.

Since  $c_1 \leq \lambda_{\min}(\mathbf{M}) \leq \lambda_{\max}(\mathbf{M}) \leq c_2$ , it follows  $\lambda_{\min} \left( \mathbf{M}'_1 \int_0^1 \mathbf{F}(t)\mathbf{F}'(t) dt \mathbf{M}'_1 \right) \geq 1/k$  in probability. Further, for any given  $j \geq 0$ ,

$$\begin{aligned} & \left\| \frac{1}{n} \sum_{t=1}^{n-j} (\mathbf{x}_{t+j,2} - \bar{\mathbf{x}}_2)(\mathbf{x}_{t2} - \bar{\mathbf{x}}_2)' - \text{Cov}(\mathbf{x}_{t+j,2}, \mathbf{x}_{t2}) \right\|_2 \\ &= \left\| \mathbf{M}'_2 \left( \frac{1}{n} \sum_{t=1}^n [(\mathbf{v}_{t+j} - \bar{\mathbf{v}})(\mathbf{v}_t - \bar{\mathbf{v}})' - \text{Cov}(\mathbf{v}_{t+j}, \mathbf{v}_t)] \right) \mathbf{M}_2 \right\|_2 = o_p(1). \end{aligned} \quad (\text{S0.10})$$

$$\begin{aligned} \left\| \frac{1}{n^{3/2}} \sum_{t=1}^{n-j} (\mathbf{x}_{t+j,1} - \bar{\mathbf{x}}_2)(\mathbf{x}_{t2} - \bar{\mathbf{x}}_2)' \right\|_2 &= \left\| \mathbf{M}'_1 \left( \frac{1}{n^{3/2}} \sum_{t=1}^n (\boldsymbol{\varsigma}_{t+j} - \bar{\boldsymbol{\varsigma}})(\mathbf{v}_t - \bar{\mathbf{v}})' \right) \mathbf{M}_2 \right\|_2 \\ &= O_p(k/n^{1/2}), \end{aligned} \quad (\text{S0.11})$$

where  $\mathbf{v}_t$  is given in (i) of Condition 3.

By (S0.9)–(S0.11), similar to the proof of Theorem 3 in ZRY, it can be shown that when  $k = o(n^{1/2-1/\eta})$ ,

$$\|\widehat{\mathbf{A}}_2 - \mathbf{A}_2\|_2 = O_p(p^{1/2}k/n). \quad (\text{S0.12})$$

Similar to (S0.9), there exists a  $k$ -dimensional Gaussian process  $\mathbf{w}(t)$  such that

$$\left\| \frac{1}{n^2} \sum_{t=1}^n \mathbf{y}_t \mathbf{y}'_t - \mathbf{A}_1 \mathbf{M}'_1 \int_0^1 \mathbf{w}(t)\mathbf{w}'(t) dt \mathbf{M}_1 \mathbf{A}'_1 \right\|_2 = o_p(1) \quad (\text{S0.13})$$

and similar to (S0.11), we can show (S0.6) holds provided  $k/n^{1/2} \rightarrow 0$  as  $n \rightarrow \infty$ . Thus, by (S0.12) and (S0.13), we also have (S0.8) and complete the proof of Lemma 5.  $\square$

**Lemma 6.** *Under Condition 1,*

$$\left\| \frac{1}{\sqrt{n}} \sum_{t=1}^n \nabla_{\mathbf{y}_t \mathbf{y}'_{t-1}} (\widehat{\mathbf{A}}_2 - \mathbf{A}_2) \right\|_2 = o_p(1),$$

and under the conditions of Theorem 3,

$$\left\| \frac{1}{\sqrt{n}} \sum_{t=1}^n \nabla_{\mathbf{y}_t \mathbf{y}'_{t-1}} (\widehat{\mathbf{A}}_2 - \mathbf{A}_2) \right\|_2 = O_p(p^{1/2} k^2 / n^{1/2}). \quad (\text{S0.14})$$

*Proof.* When  $p$  is fixed, similar to (S0.6), we have

$$\frac{1}{n^{3/2}} \left\| \sum_{t=1}^n \nabla_{\mathbf{y}_t \mathbf{y}'_{t-1}} \right\|_2 = o_p(1).$$

As a result, it follows from (S0.7) that

$$\left\| \frac{1}{\sqrt{n}} \sum_{t=1}^n \nabla_{\mathbf{y}_t \mathbf{y}'_{t-1}} (\widehat{\mathbf{A}}_2 - \mathbf{A}_2) \right\|_2 = o_p(1). \quad (\text{S0.15})$$

When  $p$  tends to infinity as  $n \rightarrow \infty$ , using the same idea as in (S0.11), we can show

$$\frac{1}{n^{3/2}} \left\| \sum_{t=1}^n \nabla_{\mathbf{y}_t \mathbf{y}'_{t-1}} \right\|_2 = O_p(k/n^{1/2}). \quad (\text{S0.16})$$

Thus, by (S0.12) and  $p \leq k = o(n^{1/2})$ , it follows that

$$\left\| \frac{1}{\sqrt{n}} \sum_{t=1}^n \nabla_{\mathbf{y}_t \mathbf{y}'_{t-1}} (\widehat{\mathbf{A}}_2 - \mathbf{A}_2) \right\|_2 = O_p(p^{1/2} k^2 / n^{1/2}).$$

Thus, we have Lemma 6.  $\square$

**Lemma 7.** *Let  $\Sigma = \text{E}\{(\mathbf{f}'_{t-1}, \dots, \mathbf{f}'_{t-s})'(\mathbf{f}'_{t-1}, \dots, \mathbf{f}'_{t-s})\}$ . Under Condition 1*

, for any given positive integer  $s$ ,

$$\frac{1}{n} \left[ \sum_{t=s+1}^n (\widehat{\mathbf{f}}'_{t-1}, \dots, \widehat{\mathbf{f}}'_{t-s})' (\widehat{\mathbf{f}}'_{t-1}, \dots, \widehat{\mathbf{f}}'_{t-s}) - M \right] \xrightarrow{p} \boldsymbol{\Sigma} \quad (\text{S0.17})$$

and under the condition of Theorem 3, in probability

$$\frac{1}{n} \left[ \sum_{t=s+1}^n (\widehat{\mathbf{f}}'_{t-1}, \dots, \widehat{\mathbf{f}}'_{t-s})' (\widehat{\mathbf{f}}'_{t-1}, \dots, \widehat{\mathbf{f}}'_{t-s}) - M \right] > 0, \quad (\text{S0.18})$$

where  $\mathbf{A} > 0$  means that  $\mathbf{A}$  is a positive definition matrix.

*Proof.* By some elementary computation, we have

$$\begin{aligned} \widehat{\mathbf{f}}_t &= [\mathbf{f}_t + \mathbf{B}'\boldsymbol{\varepsilon}_t] + [(\widehat{\mathbf{B}} - \mathbf{B})'(\mathbf{B}\mathbf{f}_t + \boldsymbol{\varepsilon}_t)] + [\widehat{\mathbf{B}}'(\mathbf{D} - \widehat{\mathbf{D}})\mathbf{x}_{t2}] + [\widehat{\mathbf{B}}'\widehat{\mathbf{D}}(\mathbf{A}_2 - \widehat{\mathbf{A}}_2)'\mathbf{y}_{t-1}] \\ &\equiv \sum_{i=1}^4 \boldsymbol{\zeta}_{t,i}. \end{aligned} \quad (\text{S0.19})$$

Next, we first show (S0.17) holds for fixed  $p$ . By (S0.33) (see below), we have

$$\|\widehat{\mathbf{B}} - \mathbf{B}\|_2 = O_p(n^{-1/2}), \quad (\text{S0.20})$$

which gives

$$\left\| \frac{1}{n} \sum_{t=s+1}^n (\boldsymbol{\zeta}'_{t-1,2}, \dots, \boldsymbol{\zeta}'_{t-s,2})' (\boldsymbol{\zeta}'_{t-1,2}, \dots, \boldsymbol{\zeta}'_{t-s,2}) \right\|_2 = o_p(1). \quad (\text{S0.21})$$

Similarly, by (S0.29) (see below) and (S0.7), we have

$$\sum_{i=3}^4 \left\| \frac{1}{n} \sum_{t=s+1}^n (\boldsymbol{\zeta}'_{t-1,i}, \dots, \boldsymbol{\zeta}'_{t-s,i})' (\boldsymbol{\zeta}'_{t-1,i}, \dots, \boldsymbol{\zeta}'_{t-s,i}) \right\|_2 = o_p(1). \quad (\text{S0.22})$$

On the other hand, by law of large numbers for  $\alpha$ -mixing process, we get

$$\frac{1}{n} \left[ \sum_{t=s+1}^n (\boldsymbol{\zeta}'_{t-1,1}, \dots, \boldsymbol{\zeta}'_{t-s,1})' (\boldsymbol{\zeta}'_{t-1,1}, \dots, \boldsymbol{\zeta}'_{t-s,1}) - M \right] \xrightarrow{p} \boldsymbol{\Sigma}. \quad (\text{S0.23})$$

Combining (S0.21)–(S0.23) yields that

$$\begin{aligned}
 & \frac{1}{n} \sum_{t=s}^n [(\widehat{\mathbf{f}}_{t-1})', \dots, (\widehat{\mathbf{f}}_{t-s})']' [(\widehat{\mathbf{f}}_{t-1})', \dots, (\widehat{\mathbf{f}}_{t-s})'] \\
 = & \frac{1}{n} \sum_{t=s+1}^n \left( \sum_{i=1}^4 \zeta'_{t-1,i}, \dots, \sum_{i=1}^4 \zeta'_{t-s,i} \right)' \left( \sum_{i=1}^4 \zeta'_{t-1,i}, \dots, \sum_{i=1}^4 \zeta'_{t-s,i} \right) \\
 = & \frac{1}{n} \sum_{t=s+1}^n (\zeta'_{t-1,1}, \dots, \zeta'_{t-s,1})' (\zeta'_{t-1,1}, \dots, \zeta'_{t-s,1}) + o_p(1) \xrightarrow{p} \Sigma
 \end{aligned}$$

and (S0.17) follows.

Now, we turn to show the case with  $p$  varying with  $n$ . Since  $p = o(n^{1/2})$ , (S0.23) still holds. Note that  $\frac{1}{n} \sum_{t=s}^n (\zeta'_{t-1,i}, \dots, \zeta'_{t-s,i})' (\zeta'_{t-1,i}, \dots, \zeta'_{t-s,i}) \geq \mathbf{0}$  for  $i = 1, \dots, 4$ . For the proof of (S0.18), it is enough to show for all  $1 \leq i \neq j \leq 4$ ,

$$\left\| \frac{1}{n} \sum_{t=s+1}^n (\zeta'_{t-1,i}, \dots, \zeta'_{t-s,i})' (\zeta'_{t-1,j}, \dots, \zeta'_{t-s,j}) \right\|_2 = o_p(1). \quad (\text{S0.24})$$

We only give  $i = 1, j = 4$  in details, other cases can be shown similarly. Since  $\mathbf{y}_t = \mathbf{A}\mathbf{x}_t$ , it follows from (2.1) that

$$\zeta_{t,1} = \mathbf{B}'(\nabla \mathbf{y}_t - \mathbf{D}\mathbf{x}_{t-1,2}) = \mathbf{B}'\mathbf{A}\mathbf{e}_t - \mathbf{B}'(\mathbf{D} + \mathbf{A}_2)\mathbf{x}_{t-1,2} = \mathbf{B}'\mathbf{A}\mathbf{M}\mathbf{v}_t - \mathbf{B}'(\mathbf{D} + \mathbf{A}_2)\mathbf{M}'_2\mathbf{v}_{t-1}.$$

Thus, by the fact that for any  $-s-1 \leq j \leq s+1$ ,

$$\left\| \sum_{t=1}^n \sum_{s=1}^t \mathbf{v}_s \mathbf{v}_{t+j}' \right\|_2 = O_p(kn) \quad (\text{S0.25})$$

and (S0.12), we have the left-hand side of (S0.24) is of order  $O_p(p^{1/2}k^2/n) = o_p(1)$ , where (S0.25) holds because the components of  $\mathbf{v}_t$  are independent. Thus, we have (S0.18) and complete the proof of Lemma 7.  $\square$

**Proof of Theorem 1.** Let  $\mathbf{b}_i, i = 1, \dots, p$  be the rows of  $\mathbf{B}$ . Lemmas 5

and 6 implies that for any  $1 \leq i \leq p$ ,

$$\begin{aligned}\sqrt{n}(\widehat{\mathbf{d}}_i - \mathbf{d}_i) &= \left( \frac{1}{\sqrt{n}} \sum_{t=1}^n (\mathbf{b}_i \mathbf{f}_t + \varepsilon_t^i) \mathbf{y}'_{t-1} \mathbf{A}_2 \right) \left( \frac{1}{n} \sum_{i=1}^n (\mathbf{A}'_2 \mathbf{y}_{t-1}) (\mathbf{A}'_2 \mathbf{y}_{t-1})' \right)^{-1} + o_p(1) \\ &= \left( \frac{1}{\sqrt{n}} \sum_{t=1}^n (\mathbf{b}_i \mathbf{f}_t + \varepsilon_t^i) \mathbf{x}'_{t-1,2} \right) \left( \frac{1}{n} \sum_{i=0}^{n-1} \mathbf{x}_{t2} \mathbf{x}'_{t2} \right)^{-1} + o_p(1).\end{aligned}\quad (\text{S0.26})$$

Since  $\{\mathbf{x}_{t2}\}$  is  $\alpha$  mixing with mixing coefficients satisfying (S0.2), it follows that

$$\frac{1}{n} \sum_{i=0}^{n-1} \mathbf{x}_{t2} \mathbf{x}'_{t2} \xrightarrow{p} \mathbb{E}(\mathbf{x}_{t2} \mathbf{x}'_{t2}) =: \mathbf{\Pi}.\quad (\text{S0.27})$$

On the other hand, by central limit theory (CLT) for  $\alpha$ -mixing process  $\{(\mathbf{b}_i \mathbf{f}_t + \varepsilon_t^i) \mathbf{x}'_{t-1,2}, 1 \leq i \leq p\}$ , there exists a  $pr \times pr$  matrix  $\mathbf{\Lambda}$  such that

$$\frac{1}{\sqrt{n}} \left( \sum_{t=1}^n (\mathbf{b}_1 \mathbf{f}_t + \varepsilon_t^1) \mathbf{x}'_{t-1,2}, \dots, \sum_{t=1}^n (\mathbf{b}_p \mathbf{f}_t + \varepsilon_t^p) \mathbf{x}'_{t-1,2} \right) \xrightarrow{d} N(0, \mathbf{\Lambda})\quad (\text{S0.28})$$

Thus, by (S0.27) and (S0.28), we have

$$\sqrt{n}(\text{vech}(\widehat{\mathbf{D}}) - \text{vech}(\mathbf{D})) \xrightarrow{d} N(0, \mathbf{\Pi}^{-1} \mathbf{\Lambda} \mathbf{\Pi}^{-1}).\quad (\text{S0.29})$$

Further, by (S0.29) and (S0.7), it is easy to show that

$$\|\widehat{\mathbf{C}} - \mathbf{C}\|_2 = \|(\widehat{\mathbf{D}} - \mathbf{D}) \mathbf{A}'_2 + \widehat{\mathbf{D}} (\widehat{\mathbf{A}}'_2 - \mathbf{A}'_2)\|_2 = O_p(n^{-1/2}).$$

Next, we show (b) of Theorem 1. Observe that

$$\widehat{\mathbf{v}}_t = \nabla \mathbf{y}_t - \widehat{\mathbf{D}} \widehat{\mathbf{A}}'_2 \mathbf{y}_{t-1} = (\nabla \mathbf{y}_t - \mathbf{D} \mathbf{x}_{t-1,2}) - (\widehat{\mathbf{D}} - \mathbf{D}) [(\widehat{\mathbf{A}}_2 - \mathbf{A}_2)' \mathbf{y}_{t-1} + \mathbf{x}_{t-1,2}] - \mathbf{D} (\widehat{\mathbf{A}}_2 - \mathbf{A}_2)' \mathbf{y}_{t-1},$$

which means that

$$\begin{aligned}
 & \frac{1}{n} \sum_{t=1}^{n-j} [\widehat{\mathbf{v}}_{t+j} \widehat{\mathbf{v}}_t' - \mathbf{E}(\nabla \mathbf{y}_{t+j} - \mathbf{D} \mathbf{x}_{t+j-1})(\nabla \mathbf{y}_t - \mathbf{D} \mathbf{x}_{t-1})'] \\
 = & \frac{1}{n} \sum_{t=1}^{n-j} [(\nabla \mathbf{y}_{t+j} - \mathbf{D} \mathbf{x}_{t+j-1})(\nabla \mathbf{y}_t - \mathbf{D} \mathbf{x}_{t-1})' - \mathbf{E}(\nabla \mathbf{y}_{t+j} - \mathbf{D} \mathbf{x}_{t+j-1})(\nabla \mathbf{y}_t - \mathbf{D} \mathbf{x}_{t-1})'] \\
 & + (\widehat{\mathbf{D}} - \mathbf{D}) \left( \frac{1}{n} \sum_{t=1}^{n-j} [(\widehat{\mathbf{A}}_2 - \mathbf{A}_2)' \mathbf{y}_{t+j-1} + \mathbf{x}_{t+j-1,2}] [(\widehat{\mathbf{A}}_2 - \mathbf{A}_2)' \mathbf{y}_{t-1} + \mathbf{x}_{t-1,2}]' \right) (\widehat{\mathbf{D}} - \mathbf{D})' \\
 & + \mathbf{D} (\widehat{\mathbf{A}}_2 - \mathbf{A}_2)' \left( \frac{1}{n} \sum_{t=1}^{n-j} \mathbf{y}_{t+j-1} \mathbf{y}_{t-1}' \right) (\widehat{\mathbf{A}}_2 - \mathbf{A}_2) \mathbf{D}' \\
 & - \frac{1}{n} \sum_{t=1}^{n-j} (\nabla \mathbf{y}_{t+j} - \mathbf{D} \mathbf{x}_{t+j-1,2}) \{ [\mathbf{y}_{t-1}' (\widehat{\mathbf{A}}_2 - \mathbf{A}_2) + \mathbf{x}_{t-1,2}'] (\widehat{\mathbf{D}} - \mathbf{D})' + \mathbf{y}_{t-1}' (\widehat{\mathbf{A}}_2 - \mathbf{A}_2) \mathbf{D}' \} \\
 & - \frac{1}{n} \sum_{t=1}^{n-j} \{ (\widehat{\mathbf{D}} - \mathbf{D}) [(\widehat{\mathbf{A}}_2 - \mathbf{A}_2)' \mathbf{y}_{t+j-1} + \mathbf{x}_{t+j-1,2}] + \mathbf{D} (\widehat{\mathbf{A}}_2 - \mathbf{A}_2)' \mathbf{y}_{t+j-1} \} (\nabla \mathbf{y}_t - \mathbf{D} \mathbf{x}_{t-1,2})' \\
 & + \frac{1}{n} \sum_{t=1}^{n-j} [(\widehat{\mathbf{A}}_2 - \mathbf{A}_2)' \mathbf{y}_{t+j-1} \mathbf{y}_{t-1}' + \mathbf{x}_{t+j-1,2} \mathbf{y}_{t-1}'] (\widehat{\mathbf{A}}_2 - \mathbf{A}_2) \mathbf{D}' \\
 & + \frac{1}{n} \sum_{t=1}^{n-j} \mathbf{D} (\widehat{\mathbf{A}}_2 - \mathbf{A}_2)' [\mathbf{y}_{t+j-1} \mathbf{y}_{t-1}' (\widehat{\mathbf{A}}_2 - \mathbf{A}_2) + \mathbf{y}_{t+j-1} \mathbf{x}_{t-1,2}'] (\widehat{\mathbf{D}} - \mathbf{D})'. \tag{S0.30}
 \end{aligned}$$

By (S0.7), (S0.29) and the law of large numbers, we have that the spectral norm of the last six terms of the right-hand side in (S0.30) is  $O_p(n^{-1})$ . And by CLT of  $\alpha$  mixing process, for any given  $j$ , the first term of the right-hand side of (S0.30) is  $O_p(n^{-1/2})$ . Similarly, we can show

$$\left\| \frac{1}{n} \sum_{t=1}^{n-j} \widehat{\mathbf{v}} \widehat{\mathbf{v}}_t' \right\|_2 = O_p(n^{-1}). \tag{S0.31}$$

Thus,

$$\|\widehat{\boldsymbol{\Sigma}}_v(j) - \boldsymbol{\Sigma}_v(j)\|_2 = O_p(n^{-1/2}), \tag{S0.32}$$

where  $\boldsymbol{\Sigma}_v(j) = \mathbf{E}(\nabla \mathbf{y}_{t+j} - \mathbf{D} \mathbf{x}_{t+j-1})(\nabla \mathbf{y}_t - \mathbf{D} \mathbf{x}_{t-1})'$ . Since  $j_0$  is fixed, it



follows from (S0.32) that

$$\|\widehat{\mathbf{W}} - \sum_{j=1}^{j_0} \boldsymbol{\Sigma}_v(j) \boldsymbol{\Sigma}'_v(j)\|_2 = O_p(n^{-1/2}). \quad (\text{S0.33})$$

Note that  $D(\mathcal{M}(\widehat{\mathbf{B}}), \mathcal{M}(\mathbf{B})) = O_p(\|\widehat{\mathbf{W}} - \sum_{j=1}^{j_0} \boldsymbol{\Sigma}_v(j) \boldsymbol{\Sigma}'_v(j)\|_2)$  (see for example, Chang, Guo and Yao (2015)), we have (b) of Theorem 1 as desired.

Now, we turn to show (c). By (S0.19), we get

$$\begin{aligned} & \sum_{t=s+1}^n [\widehat{\mathbf{f}}'_{t-1}, \dots, \widehat{\mathbf{f}}'_{t-s}]' [\widehat{\mathbf{f}}_t - \sum_{i=1}^s \mathbf{E}_i \widehat{\mathbf{f}}_{t-i}]' \\ = & \sum_{t=s+1}^n [\mathbf{f}'_{t-1} + \boldsymbol{\varepsilon}'_{t-1} \mathbf{B}, \dots, \mathbf{f}'_{t-s} + \boldsymbol{\varepsilon}'_{t-s} \mathbf{B}]' [\mathbf{e}'_t + \boldsymbol{\varepsilon}'_t \mathbf{B}] \\ & - \sum_{t=s+1}^n [\mathbf{f}'_{t-1} + \boldsymbol{\varepsilon}'_{t-1} \mathbf{B}, \dots, \mathbf{f}'_{t-s} + \boldsymbol{\varepsilon}'_{t-s} \mathbf{B}]' [\sum_{i=1}^s \boldsymbol{\varepsilon}'_{t-i} \mathbf{B} \mathbf{E}'_i] \\ & + \sum_{t=s+1}^n [\mathbf{f}'_{t-1} + \boldsymbol{\varepsilon}'_{t-1} \mathbf{B}, \dots, \mathbf{f}'_{t-s} + \boldsymbol{\varepsilon}'_{t-s} \mathbf{B}]' [\sum_{j=2}^4 (\boldsymbol{\zeta}_{t,j} - \sum_{i=1} \mathbf{E}_i \boldsymbol{\zeta}_{t-i,j})]' \\ & + \sum_{t=s+1}^n \sum_{j=2}^4 [\boldsymbol{\zeta}'_{t-1,j}, \dots, \boldsymbol{\zeta}'_{t-s,j}]' [\mathbf{e}_t + \mathbf{B}' \boldsymbol{\varepsilon}_t - \sum_{i=1}^s \mathbf{E}_i \mathbf{B}' \boldsymbol{\varepsilon}_{t-i} + \sum_{j=2}^4 (\boldsymbol{\zeta}_{t,j} - \sum_{i=1} \mathbf{E}_i \boldsymbol{\zeta}_{t-i,j})]' \\ =: & \sum_{i=1}^4 \Delta_{ni}. \end{aligned} \quad (\text{S0.34})$$

By (S0.7), (S0.20) and (S0.29), we can show that for any given positive integer  $s$ ,

$$\|\Delta_{n3}\|_2 + \|\Delta_{n4}\|_2 = O_p(\sqrt{n}). \quad (\text{S0.35})$$

On the other hand, since for any  $1 \leq i, j \leq s$  and  $l \neq i$ ,  $\text{vech}\{(\mathbf{f}_{t-i} + \mathbf{B} \boldsymbol{\varepsilon}_{t-i})(\mathbf{e}'_t + \boldsymbol{\varepsilon}'_t \mathbf{B}), \mathbf{f}_{t-i} \boldsymbol{\varepsilon}'_{t-j} \mathbf{B}, \mathbf{B}' \mathbf{v}_{t-i} \boldsymbol{\varepsilon}'_{t-l} \mathbf{B}\}$  is a  $\alpha$  mixing process with finite  $2\gamma$ -moment and mixing coefficients satisfying (S0.2), it follows from the CLT of  $\alpha$  mixing process (see for example Corollary 3.2.1 of Lin and Lu)

that for some matrix  $\Gamma_1$ ,

$$\frac{1}{\sqrt{n}} \sum_{t=s+1}^n \text{vech}\{(\mathbf{f}_{t-i} + \mathbf{B}\boldsymbol{\varepsilon}_{t-i})(\mathbf{e}'_t + \boldsymbol{\varepsilon}'_t\mathbf{B}), \mathbf{f}_{t-i}\boldsymbol{\varepsilon}'_{t-j}\mathbf{B}, \mathbf{B}'\mathbf{v}_{t-i}\boldsymbol{\varepsilon}'_{t-l}\mathbf{B}\} \xrightarrow{d} N(0, \mathbf{\Gamma}) \quad (\text{S0.36})$$

Set  $\Omega = \left[ \sum_{t=s+1}^n (\widehat{\mathbf{f}}'_{t-1}, \dots, \widehat{\mathbf{f}}'_{t-s})' (\widehat{\mathbf{f}}_{t-1}, \dots, \widehat{\mathbf{f}}_{t-s}) - M \right]$ . By the definition of  $\widehat{\mathbf{E}}_i$ ,  $i = 1, 2, \dots, s$ , we have

$$\begin{pmatrix} \widehat{\mathbf{E}}'_1 - \mathbf{E}'_1 \\ \vdots \\ \widehat{\mathbf{E}}'_s - \mathbf{E}'_s \end{pmatrix} = \Omega^{-1} \left[ \begin{pmatrix} \sum_{t=s}^n \widehat{\mathbf{f}}_{t-1} (\widehat{\mathbf{f}}_t - \sum_{i=1}^s \mathbf{E}_i \widehat{\mathbf{f}}_{t-i})' \\ \vdots \\ \sum_{t=s}^n \widehat{\mathbf{f}}_{t-s} (\widehat{\mathbf{f}}_t - \sum_{i=1}^s \mathbf{E}_i \widehat{\mathbf{f}}_{t-i})' \end{pmatrix} + M \begin{pmatrix} \mathbf{E}'_1 \\ \vdots \\ \mathbf{E}'_s \end{pmatrix} \right] \quad (\text{S0.37})$$

Thus, by Lemma 7 and (S0.34)–(S0.36), we have conclusion (c) and complete the proof of Theorem 1.  $\square$

Next, we first develop bounds for the estimated eigenvalues  $\widehat{\lambda}_j$ ,  $j = 1, 2, \dots, p$ .

**Lemma 8.** *Let  $\lambda_j$ ,  $j = 1, \dots, p$  be the eigenvalues of  $\mathbf{W}_v$ . Under Condition 1 or conditions of Theorem 3,*

$$|\widehat{\lambda}_m - \lambda_m| = O_p(pn^{-1/2}) \quad \text{and} \quad |\widehat{\lambda}_{m+1}| = O_p(pn^{-1/2}). \quad (\text{S0.38})$$

*Proof.* By (b) of Theorem 1 and (b) of Theorem 3, we have for any  $1 \leq i \leq p$ ,

$$|\widehat{\lambda}_i - \lambda_i| \leq \|\widehat{\mathbf{W}}_v - \mathbf{W}_v\|_2 = O_p(pn^{-1/2}) \quad \text{and} \quad \lambda_{m+1} = \dots = \lambda_p = 0.$$

This gives Lemma 8 as desired.  $\square$

**Proof of Theorem 2.** It is enough to show that

$$\lim_{n \rightarrow \infty} P\{\tilde{m} < m\} = 0. \quad (\text{S0.39})$$

Suppose  $\tilde{m} < m$  is true, then by Lemma 8, there exists a positive constant

$c_1$  such that

$$\lim_{n \rightarrow \infty} P\{\widehat{\lambda}_{\tilde{m}+1}/\widehat{\lambda}_{\tilde{m}} \geq c_1\} = 1, \quad \text{and} \quad \lim_{n \rightarrow \infty} P\{\widehat{\lambda}_{m+1}/\widehat{\lambda}_m < c_1/2\} = 1.$$

This implies that

$$\lim_{n \rightarrow \infty} P\{\widehat{\lambda}_{\tilde{m}+1}/\widehat{\lambda}_{\tilde{m}} > \widehat{\lambda}_{m+1}/\widehat{\lambda}_m\} = 1,$$

which contradicts the definition of  $\tilde{m}$ . Thus, (S0.39) holds.  $\square$

**Proof of Theorem 3.** Since  $p = o(n^{1/2})$  and  $\{\mathbf{x}_{t2}\}$  is a  $\alpha$  mixing process with mixing coefficients satisfying (S0.2), it follows that (S0.27) also holds for this case. Further, note that for any  $1 \leq i \leq p$  and  $1 \leq j \leq r$ , applying CLT of mixing process to  $\{(\mathbf{b}_i \mathbf{f}_t + \varepsilon_t^i) x_{t-1,2}^j\}$ , which is a  $\alpha$  mixing process with coefficients satisfying (3.2), we get

$$\left| \sum_{t=1}^n (\mathbf{b}_i \mathbf{f}_t + \varepsilon_t^i) x_{t-1,2}^j \right| = O_p(\sqrt{n}),$$

which implies

$$\left\| \frac{1}{n} \sum_{t=1}^n (\mathbf{B} \mathbf{f}_t + \boldsymbol{\varepsilon}_t) \mathbf{x}'_{t-1,2} \right\|_2 = O_p(n^{-1/2}(pr)^{1/2}). \quad (\text{S0.40})$$

Thus, by Lemmas 5 and 6,

$$\begin{aligned} \|\widehat{\mathbf{D}} - \mathbf{D}\|_2 &= \left\| \left( \frac{1}{n} \sum_{t=1}^n \nabla \mathbf{y}_t \mathbf{y}'_{t-1} \widehat{\mathbf{A}}_2 \right) \left( \frac{1}{n} \sum_{i=1}^n \widehat{\mathbf{A}}_2' \mathbf{y}_{t-1} \mathbf{y}'_{t-1} \widehat{\mathbf{A}}_2 \right)^{-1} - \mathbf{D} \right\|_2 \\ &= \left\| \left( \frac{1}{n} \sum_{t=1}^n \nabla \mathbf{y}_t \mathbf{x}'_{t-1,2} \right) \left( \frac{1}{n} \sum_{i=0}^{n-1} \mathbf{x}_{t-1,2} \mathbf{x}'_{t-1,2} \right)^{-1} - \mathbf{D} \right\|_2 + O_p(p^{1/2}k^2/n) \\ &= \left\| \left( \frac{1}{n} \sum_{t=1}^n (\mathbf{B} \mathbf{f}_t + \boldsymbol{\varepsilon}_t) \mathbf{x}'_{t-1,2} \right) \left( \frac{1}{n} \sum_{i=0}^{n-1} \mathbf{x}_{t-1,2} \mathbf{x}'_{t-1,2} \right)^{-1} \right\|_2 + O_p(p^{1/2}k^2/n) \\ &= O_p(n^{-1/2}(pr)^{1/2} + p^{1/2}k^2/n), \end{aligned} \quad (\text{S0.41})$$

this combining with (S0.12) yields

$$\|\widehat{\mathbf{C}} - \mathbf{C}\|_2 = \|(\widehat{\mathbf{D}} - \mathbf{D})\mathbf{A}'_2 + \widehat{\mathbf{D}}'(\widehat{\mathbf{A}}'_2 - \mathbf{A}'_2)\|_2 = O_p(n^{-1/2}(pr)^{1/2} + p^{1/2}k^2/n) \quad (\text{S0.42})$$

Thus, (a) of Theorem 3 follows from (S0.41) and (S0.42).

Next, we show (b). It is easy to see that

$$\left\| \frac{1}{n^2} \sum_{t=1}^{n-j} \mathbf{y}_{t-1} \mathbf{y}'_{t-1} \right\|_2 = O_p(p). \quad (\text{S0.43})$$

Thus, by (S0.12), (S0.41) and (iii) of Condition 3, it can be shown that  $\|\cdot\|_2$  norm of the last six terms of the right-hand side in (S0.30) are of order  $o(pn^{-1/2})$ , provided  $k = o(n^{1/2})$  and  $p = O(n^{1/4})$ . On the other hand, applying CLT of  $\alpha$  mixing process to the first term of the right-hand side of (S0.30), we get for any given  $j$ , this term is of order  $O_p(pn^{-1/2})$ . Similarly, we can show  $n^{-1} \sum_{t=1}^{n-j} \bar{\mathbf{v}} \bar{\mathbf{v}}'_t = O_p(n^{-1/2}p)$ . Thus,

$$\|\widehat{\boldsymbol{\Sigma}}_v(j) - \boldsymbol{\Sigma}_v(j)\|_2 = O_p(n^{-1/2}p). \quad (\text{S0.44})$$

Since  $j_0$  is fixed, it follows from (S0.44) that

$$\|\widehat{\mathbf{W}} - \sum_{j=1}^{j_0} \boldsymbol{\Sigma}_v(j) \boldsymbol{\Sigma}'_v(j)\|_2 = O_p(n^{-1/2}p). \quad (\text{S0.45})$$

Note that  $D(\mathcal{M}(\widehat{\mathbf{B}}), \mathcal{M}(\mathbf{B})) = O_p(\|\widehat{\mathbf{W}} - \sum_{j=1}^{j_0} \boldsymbol{\Sigma}_v(j) \boldsymbol{\Sigma}'_v(j)\|_2)$  (see for example, Chang, Guo and Yao (2015)), we have (b) of Theorem 3 as desired.

In the following, we give the proof of (c). Let  $\Delta_{ni}$ ,  $i = 1, 2, 3, 4$  be defined as in (S0.34). By conclusions (a), (b) of Theorem 3 and (S0.12), we can show that

$$\|\Delta_{n3} + \Delta_{n4}\|_2 = O_p(n^{1/2}(pr)^{1/2}[n^{-1/2}(pr)^{1/2} + p^{1/2}k^2/n + pn^{-1/2}] + p^{1/2}k^2/n) \quad (\text{S0.46})$$

On the other hand, applying CLT of  $\alpha$  mixing to the elements of  $\text{vech}\{(\mathbf{f}_{t-i} + \mathbf{B}\boldsymbol{\varepsilon}_{t-i})(\mathbf{e}'_t + \boldsymbol{\varepsilon}'_t\mathbf{B}), \mathbf{f}_{t-i}\boldsymbol{\varepsilon}'_{t-j}\mathbf{B}, \mathbf{B}'\mathbf{v}_{t-i}\boldsymbol{\varepsilon}'_{t-l}\mathbf{B}, l \neq i, 1 \leq i, j \leq s\}$ , we get

$$\|\Delta_{n1} + \Delta_{n2} - M\|_2 = O_p((pmn)^{1/2}). \quad (\text{S0.47})$$

Combining equations (S0.46)–(S0.47) with Lemma 7 and  $p = o(n^{1/2})$  yield

$$\|(\mathbf{E}_1, \dots, \mathbf{E}_s)\|_2 = O(p^{1/2}k^2n^{-1} + pm^{1/2}n^{-1/2}), \quad (\text{S0.48})$$

this gives (c) and completes the proof of Theorem 3.  $\square$

**Proof of Theorem 4.** By Lemma 8, Theorem 4 can be shown similarly as for Theorem 2. Therefore, we omit the detailed proofs.  $\square$

**Proof of Remark 1.** Since the proofs are similar, we only show the case with fixed  $p$  in details. It follows from the definition of  $\hat{m}$  that

$$\sum_{j=\hat{m}+1}^p \hat{\lambda}_j + \hat{m}\omega_n \leq \sum_{j=m}^p \hat{\lambda}_{p+1-j} + m\omega_n. \quad (\text{S0.49})$$

Suppose that  $\hat{m} > m$ , it follows from (S0.49) that

$$(\hat{m} - m)\omega_n \leq \sum_{j=m+1}^{\hat{m}} \hat{\lambda}_j \leq (\hat{m} - m)\hat{\lambda}_{m+1}. \quad (\text{S0.50})$$

Since  $\omega_n/n^{-1/2} \rightarrow \infty$ , it follows from Lemma 8 that equation (S0.50) holds with probability zero. This gives that

$$\lim_{n \rightarrow \infty} P\{\hat{m} > m\} = 0. \quad (\text{S0.51})$$

On the other hand, if  $\hat{m} < m$ , equation (S0.49) yields

$$(m - \hat{m})\hat{\lambda}_m \leq \sum_{j=\hat{m}+1}^m \hat{\lambda}_j \leq (m - \hat{m})\omega_n. \quad (\text{S0.52})$$

Lemma 8 implies  $\hat{\lambda}_m \geq \lambda_m/2 > 0$ . Thus, by (S0.52) and  $\omega_n \rightarrow 0$  as  $n \rightarrow \infty$ , we have

$$\lim_{n \rightarrow \infty} P\{\hat{m} < m\} = 0. \quad (\text{S0.53})$$

Equation (S0.51) together with (S0.53) give the consistency of  $\hat{m}$  as desired.



## References

- Lin, Z. and Lu, C. (1997). *Limit Theory on Mixing Dependent Random Variables*. Kluwer Academic Publishers, New York.
- Zhang, R. M., Robinosn, P. and Yao, Q. (2015). Identifying Cointegration by Eigenanalysis. *A Manuscript*.