

**TECHNICAL APPENDIX FOR GENERALIZED LEAST SQUARES
INFERENCE IN PANEL AND MULTILEVEL MODELS WITH SERIAL
CORRELATION AND FIXED EFFECTS**

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Let

$$C_{st} = x'_{st}\beta_1 + z'_{st}\beta_2^s + v_{st}, \quad (1)$$

or, in vector notation, $C_s = X_s\beta_1 + Z_s\beta_2^s + V_s$, where $C_s = [C_{s1}, \dots, C_{sT}]'$ is $T \times 1$, $X_s = [x_{s1}, \dots, x_{sT}]'$ is $T \times k_1$, $Z_s = [z_{s1}, \dots, z_{sT}]'$ is $T \times k_2$, and $V_s = [v_{s1}, \dots, v_{sT}]'$ is $T \times 1$. Also, let x_{sth} be the h^{th} element of x_{st} so that $x'_{st} = [x_{st1}, \dots, x_{stk_1}]$, and define z_{sth} similarly. Define $\ddot{v}_{st} = v_{st} - z'_{st}(Z'_s Z_s)^{-1} Z'_s V_s$, $\ddot{x}'_{st} = x'_{st} - z'_{st}(Z'_s Z_s)^{-1} Z'_s X_s$, $\ddot{V}_s = [\ddot{v}_{s1}, \dots, \ddot{v}_{sT}]'$, and $\ddot{X}_s = [\ddot{x}_{s1}, \dots, \ddot{x}_{sT}]'$. Let v_{st}^- be a $p \times 1$ vector with $v_{st}^- = [v_{s(t-p)}, \dots, v_{s(t-1)}]'$, and define \ddot{v}_{st}^- similarly.

Assumption 1 ($S \rightarrow \infty$, T fixed). Suppose the data are generated by model (1) and

- N1. $v_{st} = v_{st}^- \alpha + \eta_{st}$, where η_{st} is strictly stationary in t for each s , $E[\eta_{st}^2] = \sigma_\eta^2$, $E[\eta_{st}\eta_{s\tau}] = 0$ for $t \neq \tau$, and the roots of $1 - \alpha_1 z - \alpha_2 z^2 - \dots - \alpha_p z^p = 0$ have modulus greater than 1.
- N2. $\{X_s, V_s, \eta_s\}$ are iid across s . $\{Z_s\}$ are nonstochastic and identical across s .
- N3. (i) $\text{Rank}(\sum_{t=1}^T E[\ddot{x}_{st}\ddot{x}'_{st}]) = \text{Rank}(E[\ddot{X}'_s \ddot{X}_s]) = k_1$. (ii) $\text{Rank}(Z'_s Z_s) = k_2 \forall s$.
- N4. $E[V_s | X_s] = 0$, $E[V_s V'_s | X_s] = \Gamma(\alpha)$.
- N5. $E[\eta_{st}^4] = \mu_4 < \infty$ and $E[x_{sth}^4] \leq \Delta < \infty \forall s, t, h$.

Proposition 1. *If Assumption 1 is satisfied, $\hat{\alpha} \xrightarrow{p} \alpha_T(\alpha)$, where*

$$\alpha_T(\alpha) = E\left[\sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-\prime} \right]^{-1} E\left[\sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st} \right] = (\Gamma_p(\alpha) + \frac{1}{T-p} \Delta_\Gamma(\alpha))^{-1} (A(\alpha) + \frac{1}{T-p} \Delta_A(\alpha)).$$

Proposition 2. *Suppose $\alpha_T(\alpha)$ is continuously differentiable in α and that $H = D\alpha_T(\alpha)$ is invertible for all α such that N1 is satisfied, where $D\alpha_T(\alpha)$ is the derivative matrix of $\alpha_T(\alpha)$ in α . Define $\hat{\alpha}^{(\infty)} = \alpha_T^{-1}(\hat{\alpha})$. Then, if Assumption 1 is satisfied, $\hat{\alpha}^{(\infty)} - \alpha \xrightarrow{p} 0$ and $\sqrt{S}(\hat{\alpha}^{(\infty)} - \alpha) \xrightarrow{d} \frac{1}{T-p} H^{-1} (\Gamma_p(\alpha) + \frac{1}{T-p} \Delta_\Gamma(\alpha))^{-1} N(0, \Xi)$, where*

$$\Xi = E\left[\sum_{t_1=p+1}^T \sum_{t_2=p+1}^T \ddot{v}_{st_1}^- \ddot{\mu}_{st_1} \ddot{\mu}_{st_2} \ddot{v}_{st_2}^{-\prime} \right]$$

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and $\ddot{\mu}_{st} = \ddot{v}_{st} - \ddot{v}_{st}' \alpha_T(\alpha)$.

Assumption 2 ($S, T \rightarrow \infty$). Suppose the data are generated by model (1) and

- NT1. $v_{st} = v_{st}' \alpha + \eta_{st}$, where η_{st} is strictly stationary in t for each s , $E[\eta_{st}^2] = \sigma_\eta^2$, $E[\eta_{st}\eta_{s\tau}] = 0$ for $t \neq \tau$, and the roots of $1 - \alpha_1 z - \alpha_2 z^2 - \dots - \alpha_p z^p = 0$ have modulus greater than 1.
- NT2. $\{X_s, V_s, \eta_s\}$ are iid across s . $\{Z_s\}$ are nonstochastic and identical across s .
- NT3. (i) $[X, Z]$, where $X = [X_1', \dots, X_S']'$ and $Z = \text{diag}(Z_1, \dots, Z_S)$ has full rank. (ii) $Z_s' Z_s$ is uniformly positive definite with minimum eigenvalue $\lambda_s \geq \lambda > 0$ for all s .
- NT4. $E[V_s | X_s] = 0$, $E[V_s V_s' | X_s] = \Gamma(\alpha)$.
- NT5. $\{X_{st}, V_{st}, \eta_{st}\}$ is α -mixing of size $\frac{-3r}{r-4}$, $r > 4$, and $z_{ith}^2 \leq \Delta < \infty$, $E|x_{ith}^2|^{r+\delta} \leq \Delta < \infty$, and $E|\eta_{it}^2|^{r+\delta} \leq \Delta < \infty$ for some $\delta > 0$ and all i, t, h .

Proposition 3. *If Assumption 2 is satisfied, $\sqrt{ST}(\hat{\alpha} - \alpha) \xrightarrow{d} N(\rho B(\alpha), \Gamma_p^{-1} \Xi \Gamma_p^{-1})$ if $\frac{S}{T} \rightarrow \rho \geq 0$, where $\Xi = \lim_{T \rightarrow \infty} \frac{1}{T-p} \sum_{t_1=p+1}^T \sum_{t_2=p+1}^T E[v_{st_1}^- \eta_{st_1} \eta_{st_2} v_{st_2}^-]$. In addition, if η_{st} are independent for all s and t , $\Xi = \sigma_\eta^2 \Gamma_p$.*

Assumption 3. For $B(\alpha, T)$ defined in Lemma C.5,

- NT6. $B(\alpha, T)$ is continuously differentiable in α with bounded derivative uniformly in T for all α satisfying the stationarity condition given in NT1.

Proposition 4. *If Assumption 2 is satisfied, $\sqrt{ST}(\hat{\alpha}^{(1)} - \alpha) \xrightarrow{d} N(0, \Gamma_p^{-1} \Xi \Gamma_p^{-1})$ for Ξ defined in Proposition 3 if $\frac{S}{T} \rightarrow \rho \geq 0$. If Assumption 3 is also satisfied, then $\sqrt{ST}(\hat{\alpha}^{(1)} - \alpha) \xrightarrow{d} N(0, \Gamma_p^{-1} \Xi \Gamma_p^{-1})$ if $\frac{S}{T^3} \rightarrow 0$.*

Proposition 5. *If Assumptions 2 and 3 are satisfied and $H = D\alpha_T(\alpha)$ is invertible for all α satisfying the stationarity condition given in NT1 uniformly in T , where $D\alpha_T(\alpha)$ is the derivative matrix of $\alpha_T(\alpha)$ in α , $\sqrt{ST}(\hat{\alpha}^{(\infty)} - \alpha) \xrightarrow{d} N(0, \Gamma_p^{-1} \Xi \Gamma_p^{-1})$ for Ξ defined in Proposition 3.*

Assumption 4 (Higher-Order Asymptotics). Suppose the data are generated by model (1) and

- HO1. $v_{st} = v_{st}' \alpha + \eta_{st}$, where η_{st} are iid $N(0, \sigma_\eta^2)$ random variables, and $E[V_s V_s'] = \Gamma(\alpha)$ is a $T \times T$ positive definite matrix with minimum eigenvalue bounded away from 0 and maximum eigenvalue bounded away from infinity uniformly in T .
- HO2. (i) $\{X_s, Z_s\}$ are nonstochastic with $\{Z_s\}$ identical across s . (ii) $[X, Z]$, where $X = [X_1', \dots, X_S']'$ and $Z = \text{diag}(Z_1, \dots, Z_S)$ has full rank, and the minimum eigenvalue of $X'X - X'Z(Z'Z)^{-1}Z'X$ is bounded away from zero. (iii) $|x_{sth}| \leq \Delta$ and $|z_{sth}| \leq \Delta$.

- HO3. Let $M_z(\tilde{\alpha}) = (I \otimes \Gamma(\tilde{\alpha})^{-1}) - (I \otimes \Gamma(\tilde{\alpha})^{-1})Z(Z'(I \otimes \Gamma(\tilde{\alpha})^{-1})Z)^{-1}Z'(I \otimes \Gamma(\tilde{\alpha})^{-1})$, $M = I - X(X'M_z(\alpha)X)^{-1}X'M_z(\alpha)$, $A(\tilde{\alpha}) = \frac{1}{ST}X'M_z(\tilde{\alpha})X$, and $\Psi(\tilde{\alpha}) = \frac{1}{\sqrt{ST}}X'M_z(\tilde{\alpha})MV$. For $i = 1, \dots, p$, $j = 1, \dots, p$, and $k = 1, \dots, p$, (i) each matrix in $A(\alpha)$, $A_i(\alpha) = \frac{\partial A}{\partial \alpha_i}|_{\alpha}$, and $A_{ij} = \frac{\partial^2 A}{\partial \alpha_i \partial \alpha_j}|_{\alpha}$ approaches a limit as $S, T \rightarrow \infty$, and $\lim A(\alpha)$ is nonsingular, (ii) the covariance matrices for the vectors in $\Psi_i(\alpha) = \frac{\partial \Psi}{\partial \alpha_i}|_{\alpha}$, and $\Psi_{ij}(\alpha) = \frac{\partial^2 \Psi}{\partial \alpha_i \partial \alpha_j}|_{\alpha}$ approach limits as $S, T \rightarrow \infty$, and (iii) all the matrices in $A(\tilde{\alpha})$, $A_i(\tilde{\alpha})$, $A_{ij}(\tilde{\alpha})$, $A_{ijk}(\tilde{\alpha}) = \frac{\partial^3 A}{\partial \alpha_i \partial \alpha_j \partial \alpha_k}|_{\tilde{\alpha}}$, $\Psi(\tilde{\alpha})$, $\Psi_i(\tilde{\alpha})$, $\Psi_{ij}(\tilde{\alpha})$, and $\Psi_{ijk}(\tilde{\alpha}) = \frac{\partial^3 \Psi}{\partial \alpha_i \partial \alpha_j \partial \alpha_k}|_{\tilde{\alpha}}$ are bounded in probability uniformly for $\tilde{\alpha} \in \mathcal{A}$ where \mathcal{A} is a neighborhood of α .
- HO4. $B(\alpha, T)$ is three times continuously differentiable in α with the first three derivatives bounded uniformly in T for all α satisfying the stationarity condition given in NT1.

Proposition 6. Let $\hat{\alpha}$ be the least squares estimator of α , $\hat{\alpha}^{(\infty)}$ be the iteratively bias-corrected estimator, and $\hat{\beta}(\hat{\alpha})$ and $\hat{\beta}(\hat{\alpha}^{(\infty)})$ be the corresponding FGLS estimators. If Assumption 4 is satisfied and $\frac{S}{T} \rightarrow \rho$ as $\{S, T\} \rightarrow \infty$ jointly, the higher-order bias and variance of $\hat{\beta}(\hat{\alpha})$ and $\hat{\beta}(\hat{\alpha}^{(\infty)})$ are

$$\text{Bias}(\hat{\beta}(\hat{\alpha})) = \text{Bias}(\hat{\beta}(\hat{\alpha}^{(\infty)})) = 0 \quad (2)$$

$$\text{Var}(\sqrt{ST}(\hat{\beta}(\hat{\alpha}) - \beta)) = A^{-1} + \Upsilon/ST \quad (3)$$

$$+ \frac{1}{ST} \sum_{i=1}^p \sum_{j=1}^p \zeta_{ij} \frac{S}{T} B(\alpha, T)_i B(\alpha, T)_j + O(1/ST^2)$$

$$\text{Var}(\sqrt{ST}(\hat{\beta}(\hat{\alpha}^{(\infty)}) - \beta)) = A^{-1} + \Upsilon/ST + O(1/ST^2). \quad (4)$$

for $\zeta_{ij} = E[\Psi_i \Psi_j']$ and $\Upsilon = \sum_{i=1}^p \sum_{j=1}^p \zeta_{ij} (\frac{ST}{S(T-p)} \Gamma_p^{-1})_{[i,j]}$ where $(\frac{ST}{S(T-p)} \Gamma_p^{-1})_{[i,j]}$ is the $[i, j]$ element of matrix $\frac{ST}{S(T-p)} \Gamma_p^{-1}$. Also,

$$\begin{aligned} & \text{Var}(\sqrt{ST}(\hat{\beta}(\hat{\alpha}) - \beta)) - \text{Var}(\sqrt{ST}(\hat{\beta}(\hat{\alpha}^{(\infty)}) - \beta)) \\ &= \frac{1}{ST} \sum_{i=1}^p \sum_{j=1}^p \zeta_{ij} \frac{S}{T} B(\alpha, T)_i B(\alpha, T)_j \geq 0. \end{aligned}$$

APPENDIX A. NOTATION

The following notation will be used throughout the appendix.

Suppose the panel has a cross-sectional dimension $s = 1, \dots, S$ and a time-series dimension $t = 1, \dots, T$ with $T > 2p$ where p is the order of the autoregressive process. Let

$$C_{st} = x'_{st} \beta_1 + z'_{st} \beta_2^s + v_{st}, \quad (5)$$

or, in vector notation, $C_s = X_s\beta_1 + Z_s\beta_2^s + V_s$, where $C_s = [C_{s1}, \dots, C_{sT}]'$ is $T \times 1$, $X_s = [x_{s1}, \dots, x_{sT}]'$ is $T \times k_1$, $Z_s = [z_{s1}, \dots, z_{sT}]'$ is $T \times k_2$, and $V_s = [v_{s1}, \dots, v_{sT}]'$ is $T \times 1$. Also, let x_{sth} be the h^{th} element of x_{st} so that $x'_{st} = [x_{st1}, \dots, x_{stk_1}]$, and define z_{sth} similarly.

Let v_{st}^- be a $p \times 1$ vector with $v_{st}^- = [v_{s(t-p)}, \dots, v_{s(t-1)}]'$.

Define $\ddot{v}_{st} = v_{st} - z'_{st}(Z'_s Z_s)^{-1} Z'_s V_s$, $\ddot{x}'_{st} = x'_{st} - z'_{st}(Z'_s Z_s)^{-1} Z'_s X_s$, $\ddot{V}_s = [\ddot{v}_{s1}, \dots, \ddot{v}_{sT}]'$, and $\ddot{X}_s = [\ddot{x}_{s1}, \dots, \ddot{x}_{sT}]'$.

Throughout, let $\|A\| = [\text{trace}(A'A)]^{1/2}$ be the Euclidean norm of a matrix A .

APPENDIX B. PROOF OF PROPOSITION 1 AND PROPOSITION 2

Proposition 1 is verified by combining Lemmas B.3 and B.4 below, and Proposition 2 follows from Proposition 1 and Lemma B.6.

All results presented below are for asymptotics where $S \rightarrow \infty$ with T fixed.

Proof of Proposition 1. Immediate from Lemma B.3 and Lemma B.4. ■

Proof of Proposition 2. That $\alpha_T(\alpha)$ is continuously differentiable in α and that $H = D\alpha_T(\alpha)$ is invertible for all α such that N1 is satisfied imply that $\alpha_T(\alpha)$ is invertible for all α such that N1 is satisfied by the Inverse Function Theorem. (See, e.g. Fitzpatrick (1996) Theorem 16.9.) $\hat{\alpha}^{(\infty)} - \alpha \xrightarrow{p} 0$ then follows immediately from the definition of $\hat{\alpha}^{(\infty)}$ and Proposition 1.

To verify the asymptotic normality, expand $\hat{\alpha}^{(\infty)}$ about $\hat{\alpha} = \alpha_T(\alpha)$. This gives

$$\hat{\alpha}^{(\infty)} = \alpha_T^{-1}(\alpha_T(\alpha)) + H^{-1} \widetilde{\big|}_{\alpha_T(\alpha)} (\hat{\alpha} - \alpha_T(\alpha)),$$

where $\widetilde{\big|}_{\alpha_T(\alpha)}$ is an intermediate value between $\hat{\alpha}$ and $\alpha_T(\alpha)$. The conclusion then follows from continuity of H , Proposition 1, and Lemma B.6. ■

B.1. Lemmas.

Lemma B.1. Let $\hat{\beta}_1$ be the ordinary least squares estimate of β_1 . Then if the conditions of Assumption 1 are satisfied, $\hat{\beta}_1 - \beta_1 \xrightarrow{p} 0$ and $\sqrt{S}(\hat{\beta}_1 - \beta_1) \xrightarrow{d} N(0, M^{-1}\Omega M^{-1})$, where $M = E[\ddot{X}'_s \ddot{X}_s]$ and $\Omega = E[\ddot{X}'_s \Gamma(\alpha) \ddot{X}_s]$.

Proof. Using $\|AB\| \leq \|A\| \|B\|$ and $I - Z_s(Z'_s Z_s)^{-1} Z'_s$ positive semi-definite, $E\|\ddot{X}'_s \ddot{X}_s\| \leq E\|\ddot{X}_s\|^2 = E[\text{trace}(X'_s X_s - X'_s Z_s(Z'_s Z_s)^{-1} Z'_s X_s)] \leq \text{trace}(E X'_s X_s) = \text{trace}(\sum_{t=1}^T E[x_{st} x'_{st}]) < \infty$ by N5. Also, $\|\ddot{X}'_s \ddot{V}_s\| \leq (E\|\ddot{X}'_s\|^2 E\|\ddot{V}_s\|^2)^{1/2} < \infty$ from the Cauchy-Schwarz inequality, N5, and the same argument as above. The Khinchin LLN then yields $\frac{1}{S} \sum_{s=1}^S \ddot{X}'_s \ddot{X}_s \xrightarrow{p} M$ and $\frac{1}{S} \sum_{s=1}^S \ddot{X}'_s \ddot{V}_s \xrightarrow{p} 0$, from which $\hat{\beta}_1 - \beta_1 \xrightarrow{p} 0$ follows.

To show asymptotic normality of $\widehat{\beta}_1$ first note that $\ddot{X}'_s \ddot{V}_s$ is iid and has mean zero by N2 and N4. Also, $\mathbb{E}\|\ddot{X}'_s \ddot{V}_s \ddot{V}'_s \ddot{X}_s\| \leq (2\mathbb{E}\|X_s\|^4 \mathbb{E}\|V_s\|^4)^{1/2} < \infty$ by N5, $\|AB\| \leq \|A\|\|B\|$, the Cauchy-Schwarz inequality, and $\mathbb{E}\|\ddot{X}_s\|^4 = \mathbb{E}[(\text{trace}(X'_s X_s))^2 - 2\text{trace}(X'_s X_s)\text{trace}(X'_s Z_s (Z'_s Z_s)^{-1} Z'_s X_s) + (\text{trace}(X'_s Z_s (Z'_s Z_s)^{-1} Z'_s X_s))^2] \leq \mathbb{E}[2(\text{trace}(X'_s X_s))^2] = 2\mathbb{E}\|X_s\|^4$, where the inequality follows from $X'_s X_s$, $X'_s Z_s (Z'_s Z_s)^{-1} Z'_s X_s$, and $I - Z_s (Z'_s Z_s)^{-1} Z'_s$ positive semi-definite. It then follows from the Lindeberg-Levy CLT that $\frac{1}{\sqrt{S}} \sum_{s=1}^S \ddot{X}'_s \ddot{V}_s \xrightarrow{d} N(0, \Omega)$ since $\mathbb{E}[\ddot{X}'_s \ddot{V}_s \ddot{V}'_s \ddot{X}_s] = \mathbb{E}[\ddot{X}'_s V_s V'_s \ddot{X}_s] = \mathbb{E}[\ddot{X}'_s \Gamma(\alpha) \ddot{X}_s]$, from which $\sqrt{S}(\widehat{\beta}_1 - \beta_1) \xrightarrow{d} N(0, M^{-1} \Omega M^{-1})$ is obtained. ■

Lemma B.2. Define \tilde{v}_{st} to be the residual from least squares regression of (1); i.e. $\tilde{v}_{st} = C_{st} - x'_{st} \widehat{\beta}_1 - z'_{st} \widehat{\beta}_2^s = v_{st} - x'_{st}(\widehat{\beta}_1 - \beta_1) - z'_{st}(\widehat{\beta}_2^s - \beta_2) = \ddot{v}_{st} - \ddot{x}'_{st}(\widehat{\beta}_1 - \beta_1)$, where $\widehat{\beta}_1$ and $\widehat{\beta}_2^s$ are least squares estimates of β_1 and β_2^s . Let \tilde{v}_{st}^- be a $p \times 1$ vector with $\tilde{v}_{st}^- = [\tilde{v}_{s(t-p)}, \dots, \tilde{v}_{s(t-1)}]'$, and let \ddot{v}_{st}^- be a $p \times 1$ vector with $\ddot{v}_{st}^- = [\ddot{v}_{s(t-p)}, \dots, \ddot{v}_{s(t-1)}]'$. Under the conditions of Assumption 1, $\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \tilde{v}_{st}^- \tilde{v}_{st}^{-\prime} = \frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-\prime} + o_p(S^{-1/2})$, and $\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \tilde{v}_{st}^- \tilde{v}_{st}^- = \frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^- + o_p(S^{-1/2})$.

Proof.

$$\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \tilde{v}_{st}^- \tilde{v}_{st}^- = \frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T (\ddot{v}_{st}^- \ddot{v}_{st}^- - \ddot{x}_{st}^- (\widehat{\beta}_1 - \beta_1) \ddot{v}_{st}^- - \ddot{v}_{st}^- \ddot{x}'_{st} (\widehat{\beta}_1 - \beta_1) + \ddot{x}_{st}^- (\widehat{\beta}_1 - \beta_1) (\widehat{\beta}_1 - \beta_1)' \ddot{x}_{st}^-),$$

where \ddot{x}_{st}^- is a $p \times k_1$ matrix with $\ddot{x}_{st}^- = [\ddot{x}_{s(t-p)}, \dots, \ddot{x}_{s(t-1)}]'$. Note

$$\text{vec}\left(\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{x}_{st}^- (\widehat{\beta}_1 - \beta_1) \ddot{v}_{st}^-\right) = \left\{ \frac{1}{S} \sum_{s=1}^S \left[\sum_{t=p+1}^T (\ddot{v}_{st}^- \otimes \ddot{x}_{st}^-) \right] \right\} (\widehat{\beta}_1 - \beta_1) = o_p(1) O_p(S^{-1/2})$$

by Lemma B.1, Assumption 1, and the Khinchin LLN. Similarly,

$$\text{vec}\left(\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{x}'_{st} (\widehat{\beta}_1 - \beta_1)\right) = o_p(1) O_p(S^{-1/2}),$$

and

$$\text{vec}\left(\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{x}_{st}^- (\widehat{\beta}_1 - \beta_1) (\widehat{\beta}_1 - \beta_1)' \ddot{x}_{st}^-\right) = O_p(1) O_p(S^{-1/2}) O_p(S^{-1/2}).$$

It then follows that $\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \tilde{v}_{st}^- \tilde{v}_{st}^- = \frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^- + o_p(S^{-1/2})$ and, by a similar argument, $\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \tilde{v}_{st}^- \tilde{v}_{st}^{-\prime} = \frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-\prime} + o_p(S^{-1/2})$. ■

Lemma B.3. Under the conditions of Assumption 1, $\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-\prime} \xrightarrow{p} \mathbb{E}[\sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-\prime}] = (T-p)(\Gamma_p(\alpha) + \frac{1}{T-p} \Delta_\Gamma(\alpha))$, and $\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^- \xrightarrow{p} \mathbb{E}[\sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^-] = (T-p)(A(\alpha) + \frac{1}{T-p} \Delta_A(\alpha))$, where $\Gamma_p(\alpha) + \frac{1}{T-p} \Delta_\Gamma(\alpha)$ is a $p \times p$ matrix with i, j element

$$\begin{aligned} \mathbb{E} \left[\frac{1}{T-p} \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-\prime} \right]_{[i,j]} &= \gamma_{|i-j|}(\alpha) - \frac{1}{T-p} \text{trace}(Z'_s \Gamma_{-i}(\alpha) Z_{s,-j} (Z'_s Z_s)^{-1}) \\ &- \frac{1}{T-p} \text{trace}(Z'_s \Gamma_{-j}(\alpha) Z_{s,-i} (Z'_s Z_s)^{-1}) \\ &+ \frac{1}{T-p} \text{trace}(Z'_s \Gamma(\alpha) Z_s (Z'_s Z_s)^{-1} Z'_{s,-i} Z_{s,-j} (Z'_s Z_s)^{-1}), \end{aligned} \quad (6)$$

$A(\alpha) + \frac{1}{T-p}\Delta_A(\alpha)$ is a $p \times 1$ vector with i^{th} element

$$\begin{aligned} \mathbb{E} \left[\frac{1}{T-p} \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^- \right]_{[i,1]} &= \gamma_i(\alpha) - \frac{1}{T-p} \text{trace} (Z'_s \Gamma_{-i}(\alpha) Z_{s,-0} (Z'_s Z_s)^{-1}) \\ &- \frac{1}{T-p} \text{trace} (Z'_s \Gamma_{-0}(\alpha) Z_{s,-i} (Z'_s Z_s)^{-1}) \\ &+ \frac{1}{T-p} \text{trace} (Z'_s \Gamma(\alpha) Z_s (Z'_s Z_s)^{-1} Z'_{s,-i} Z_{s,-0} (Z'_s Z_s)^{-1}), \end{aligned} \quad (7)$$

$\Gamma(\alpha) = \mathbb{E}[V_s V'_s]$, $\Gamma_{-k}(\alpha) = \mathbb{E}[V_s V'_{s,-k}]$, $\gamma_i(\alpha) = \mathbb{E}[v_{st} v_{s(t-i)}]$, $V_{s,-k} = (v_{s(p+1-k)}, v_{s(p+2-k)}, \dots, v_{s(T-k)})'$ and $Z_{s,-k}$ is defined similarly.

Proof. The proof is given for $\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^- \xrightarrow{p} \mathbb{E}[\sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^-] = (T-p)(\Gamma_p(\alpha) + \frac{1}{T-p}\Delta\Gamma(\alpha))$; $\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^- \xrightarrow{p} \mathbb{E}[\sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^-] = (T-p)(A(\alpha) + \frac{1}{T-p}\Delta_A(\alpha))$ follows by a similar argument.

$\frac{1}{S} \sum_{s=1}^S \frac{1}{T-p} \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^-$ is a $p \times p$ matrix with $[i, j]$ element

$$\begin{aligned} \left[\frac{1}{S} \sum_{s=1}^S \frac{1}{T-p} \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^- \right]_{[i,j]} &= \frac{1}{S} \sum_{s=1}^S \frac{1}{T-p} \sum_{t=p+1}^T v_{s(t-i)} v_{s(t-j)} \\ &- \frac{1}{S} \sum_{s=1}^S \frac{1}{T-p} \sum_{t=p+1}^T v_{s(t-i)} z'_{s(t-j)} (Z'_s Z_s)^{-1} Z'_s V_s \\ &- \frac{1}{S} \sum_{s=1}^S \frac{1}{T-p} \sum_{t=p+1}^T v_{s(t-j)} z'_{s(t-i)} (Z'_s Z_s)^{-1} Z'_s V_s \\ &+ \frac{1}{S} \sum_{s=1}^S \frac{1}{T-p} \sum_{t=p+1}^T z'_{s(t-i)} (Z'_s Z_s)^{-1} Z'_s V_s V'_s Z_s (Z'_s Z_s)^{-1} z_{s(t-j)} \\ &= \frac{1}{S} \sum_{s=1}^S \frac{1}{T-p} \sum_{t=p+1}^T v_{s(t-i)} v_{s(t-j)} \\ &- \frac{1}{S} \sum_{s=1}^S \frac{1}{T-p} V'_{s,-i} Z_{s,-j} (Z'_s Z_s)^{-1} Z'_s V_s \\ &- \frac{1}{S} \sum_{s=1}^S \frac{1}{T-p} V'_{s,-j} Z_{s,-i} (Z'_s Z_s)^{-1} Z'_s V_s \\ &+ \frac{1}{S} \sum_{s=1}^S \frac{1}{T-p} Z_{s,-i} (Z'_s Z_s)^{-1} Z'_s V_s V'_s Z'_s (Z'_s Z_s)^{-1} Z'_{s,-j} \\ &\xrightarrow{p} \mathbb{E} \left[\frac{1}{T-p} \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^- \right]_{[i,j]} \end{aligned} \quad (8)$$

by the Khinchin LLN and repeated application of the triangle and Cauchy-Schwarz inequalities since

- (i) $[\frac{1}{T-p} \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-'}]_{[i,j]}$ is iid by N2,
- (ii) $\mathbb{E}|v_{s(t-i)} v_{s(t-j)}| < \infty$ by N1,
- (iii) $\mathbb{E}|v_{s(t-i)} z'_{s(t-j)} (Z'_s Z_s)^{-1} Z'_s V_s| \leq \left(\mathbb{E}|z'_{s(t-j)} (Z'_s Z_s)^{-1} Z'_s V_s|^2 \mathbb{E}|v_{s(t-i)}|^2 \right)^{1/2}$,
- (iv) $\mathbb{E}|z'_{s(t-i)} (Z'_s Z_s)^{-1} Z'_s V_s V'_s Z_s (Z'_s Z_s)^{-1} z_{s(t-j)}|$
 $\leq \left(\mathbb{E}|z'_{s(t-i)} (Z'_s Z_s)^{-1} Z'_s V_s|^2 \mathbb{E}|z'_{s(t-j)} (Z'_s Z_s)^{-1} Z'_s V_s|^2 \right)^{1/2}$, and
- (v) by N1 and N5,

$$\begin{aligned}
\mathbb{E}|z'_{s(t-j)} (Z'_s Z_s)^{-1} Z'_s V_s|^2 &= \mathbb{E}|z'_{s(t-j)} (Z'_s Z_s)^{-1} (Z'_s Z_s) (Z'_s Z_s)^{-1} Z'_s V_s|^2 \\
&\leq \|z'_{s(t-j)} (Z'_s Z_s)^{-1} Z'_s\|^2 \mathbb{E}\|Z_s (Z'_s Z_s)^{-1} Z'_s V_s\|^2 \\
&= \text{trace}(z'_{s(t-j)} (Z'_s Z_s)^{-1} z_{s(t-j)}) \times \\
&\quad \mathbb{E}[\text{trace}(V'_s Z_s (Z'_s Z_s)^{-1} Z'_s V_s)] \\
&\leq k_2 \mathbb{E}[\text{trace}(V'_s V_s)] = k_2 \mathbb{E}\left[\sum_{t=p+1}^T v_{st}^2 \right] < \infty,
\end{aligned}$$

from which it follows that $\mathbb{E}\left[\frac{1}{T-p} \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-'} \right]_{[i,j]} < \infty$. ■

Lemma B.4. Let $\hat{\alpha} = (\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-'})^{-1} (\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \tilde{v}_{st})$ be the least squares estimate of α using the least squares residuals, \tilde{v}_{st} from estimating β_1 . If Assumption 1 is satisfied,

$$\hat{\alpha} = \left(\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-'} \right)^{-1} \left(\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \tilde{v}_{st} \right) + o_p(S^{-1/2}).$$

Proof. By Lemma B.2,

$$\hat{\alpha} = \left(\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-'} + M_1 \right)^{-1} \left(\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \tilde{v}_{st} + M_2 \right),$$

where $M_1 = o_p(S^{-1/2})$ and $M_2 = o_p(S^{-1/2})$. After some algebra, it then follows that

$$\begin{aligned}
\hat{\alpha} &= \left(\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-'} \right)^{-1} \left(\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \tilde{v}_{st} \right) \\
&+ \left(\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-'} + M_1 \right)^{-1} M_2 \\
&- \left(\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-'} + M_1 \right)^{-1} M_1 \left(\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-'} \right)^{-1} \left(\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \tilde{v}_{st} \right) \\
&= \left(\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-'} \right)^{-1} \left(\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \tilde{v}_{st} \right) + O_p(1) o_p(S^{-1/2}) + O_p(1) o_p(S^{-1/2}) O_p(1) O_p(1)
\end{aligned}$$

where the last equality is by Lemma B.3, which yields the conclusion. ■

Lemma B.5. Define $\ddot{\mu}_{st} = \ddot{v}_{st} - \ddot{v}_{st}' \alpha_T(\alpha)$. If Assumption 1 is satisfied, $\frac{1}{\sqrt{S}} \sum_{s=1}^S (\sum_{t=p+1}^T \ddot{v}_{st} \ddot{\mu}_{st}) \xrightarrow{d} N(0, \Xi)$, where $\Xi = \mathbb{E}[\sum_{t_1=p+1}^T \sum_{t_2=p+1}^T \ddot{v}_{st_1} \ddot{\mu}_{st_1} \ddot{\mu}_{st_2} \ddot{v}_{st_2}']$.

Proof. $\sum_{t=p+1}^T \ddot{v}_{st} \ddot{\mu}_{st}$ are iid by N2. Also,

$$\begin{aligned} \mathbb{E}\left[\sum_{t=p+1}^T \ddot{v}_{st} \ddot{\mu}_{st}\right] &= \mathbb{E}\left[\sum_{t=p+1}^T \ddot{v}_{st} (\ddot{v}_{st} - \ddot{v}_{st}' \alpha_T(\alpha))\right] \\ &= \mathbb{E}\left[\sum_{t=p+1}^T \ddot{v}_{st} \ddot{v}_{st}\right] - \mathbb{E}\left\{\sum_{t=p+1}^T \ddot{v}_{st} \ddot{v}_{st}' (\mathbb{E}\left[\sum_{t=p+1}^T \ddot{v}_{st} \ddot{v}_{st}'\right])^{-1} \mathbb{E}\left[\sum_{t=p+1}^T \ddot{v}_{st} \ddot{v}_{st}\right]\right\} \\ &= 0, \end{aligned}$$

where the second equality comes from $\alpha_T(\alpha) = \mathbb{E}[\sum_{t=p+1}^T \ddot{v}_{st} \ddot{v}_{st}']^{-1} \mathbb{E}[\sum_{t=p+1}^T \ddot{v}_{st} \ddot{v}_{st}]$ from Proposition 1. Also

$$\begin{aligned} &\mathbb{E}\left[\sum_{t_1=p+1}^T \sum_{t_2=p+1}^T \ddot{v}_{s(t_1-i)} \ddot{\mu}_{st_1} \ddot{\mu}_{st_2} \ddot{v}_{s(t_2-j)}\right] \\ &= \mathbb{E}\left[\sum_{t_1=p+1}^T \sum_{t_2=p+1}^T \ddot{v}_{s(t_1-i)} \ddot{v}_{st_1} \ddot{v}_{st_2} \ddot{v}_{s(t_2-j)}\right] - \mathbb{E}\left[\sum_{k=1}^p \sum_{t_1=p+1}^T \sum_{t_2=p+1}^T \ddot{v}_{s(t_1-i)} \ddot{v}_{s(t_1-k)} \ddot{v}_{st_2} \ddot{v}_{s(t_2-j)} \alpha_T(\alpha)_k\right] \\ &\quad - \mathbb{E}\left[\sum_{k=1}^p \sum_{t_1=p+1}^T \sum_{t_2=p+1}^T \ddot{v}_{s(t_1-i)} \ddot{v}_{st_1} \ddot{v}_{s(t_2-k)} \ddot{v}_{s(t_2-j)} \alpha_T(\alpha)_k\right] \\ &\quad + \mathbb{E}\left[\sum_{k_1=1}^p \sum_{k_2=1}^p \sum_{t_1=p+1}^T \sum_{t_2=p+1}^T \ddot{v}_{s(t_1-i)} \ddot{v}_{s(t_1-k_1)} \ddot{v}_{s(t_2-k_2)} \ddot{v}_{s(t_2-j)} \alpha_T(\alpha)_{k_1} \alpha_T(\alpha)_{k_2}\right] \\ &< \infty \end{aligned}$$

since $\mathbb{E}|\ddot{v}_{st_1} \ddot{v}_{st_2} \ddot{v}_{st_3} \ddot{v}_{st_4}| \leq (\mathbb{E}|\ddot{v}_{st_1}|^4 \mathbb{E}|\ddot{v}_{st_2}|^4 \mathbb{E}|\ddot{v}_{st_3}|^4 \mathbb{E}|\ddot{v}_{st_4}|^4)^{1/4} < \infty$ by repeated application of the Cauchy-Schwarz inequality and $\mathbb{E}\|\ddot{v}_{st}\|^4 = \mathbb{E}\|e_t'(V_s - Z_s(Z_s'Z_s)^{-1}Z_s'V_s)\|^4 \leq \mathbb{E}(\|e_t\|^4 \|V_s\|^4) \|I_T - Z_s(Z_s'Z_s)^{-1}Z_s'\|^4 = (T - k_2)^2 \mathbb{E}\|V_s\|^4 < \infty$ by N5 and T fixed, where e_t is the t^{th} unit vector. Then the conclusion follows by the Lindeberg-Levy CLT since

$$\mathbb{E}\left[\sum_{t_1=p+1}^T \sum_{t_2=p+1}^T \ddot{v}_{st_1} \ddot{\mu}_{st_1} \ddot{\mu}_{st_2} \ddot{v}_{st_2}'\right]_{[i,j]} = \mathbb{E}\left[\sum_{t_1=p+1}^T \sum_{t_2=p+1}^T \ddot{v}_{s(t_1-i)} \ddot{\mu}_{st_1} \ddot{\mu}_{st_2} \ddot{v}_{s(t_2-j)}\right] < \infty \quad \forall i, j.$$

■

Lemma B.6. Suppose Assumption 1 holds, then $\sqrt{S}(\hat{\alpha} - \alpha_T(\alpha)) \xrightarrow{d} \frac{1}{T-p}(\Gamma_p(\alpha) + \frac{1}{T-p}\Delta_\Gamma(\alpha))^{-1}N(0, \Xi)$.

Proof. For $\ddot{\mu}_{st}$ defined in Lemma B.5, $\hat{\alpha} - \alpha_T(\alpha) = (\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st} \ddot{v}_{st}')^{-1} (\frac{1}{S} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st} \ddot{\mu}_{st}) + o_p(S^{-1/2})$ by Lemma B.4. The conclusion is then immediate from Lemmas B.3 and B.5. ■

APPENDIX C. PROOF OF PROPOSITIONS 3, 4, AND 5

The proofs of Propositions 3, 4, and 5 are collected below. All results presented below are for asymptotics where $S, T \rightarrow \infty$.

Proof of Proposition 3. $\sqrt{ST}(\hat{\alpha} - \alpha) = \sqrt{ST}(\hat{\alpha} - \alpha_T(\alpha) + \alpha_T(\alpha) - \alpha) = \sqrt{ST}(\hat{\alpha} - \alpha_T(\alpha)) + \sqrt{\frac{S}{T}}B(\alpha, T)$ by Lemma C.5, from which the conclusion follows by Lemmas C.5 and C.8. ■

Proof of Proposition 4.

$$\begin{aligned}\sqrt{ST}(\hat{\alpha}^{(1)} - \alpha) &= \sqrt{ST}[\hat{\alpha} - (-\hat{\alpha} + \alpha_T(\hat{\alpha})) - \alpha_T(\alpha) + \alpha_T(\alpha) - \alpha] \\ &= \sqrt{ST}(\hat{\alpha} - \alpha_T(\alpha)) + \sqrt{\frac{S}{T}}(B(\alpha, T) - B(\hat{\alpha}, T)).\end{aligned}$$

The first conclusion then follows from Lemmas C.5, C.8, C.9, and the Continuous Mapping Theorem, and the second conclusion follows from Lemmas C.5, C.8, C.9, and a Taylor expansion of $B(\hat{\alpha}, T)$ about $\hat{\alpha} = \alpha$. ■

Proof of Proposition 5. Recall $\hat{\alpha}^{(\infty)} = \alpha_T^{-1}(\hat{\alpha}) = \alpha_T^{-1}(\alpha_T(\alpha)) + H^{-1}|_{\alpha_T(\alpha)}(\hat{\alpha} - \alpha_T(\alpha))$. From Lemma C.5, $\alpha_T(\alpha) = \alpha + \frac{1}{T-p}B(\alpha, T)$ which implies $H^{-1} \rightarrow I$ as $T \rightarrow \infty$ by NT6. The conclusion then follows from Lemma C.8. ■

C.1. Lemmas.

Lemma C.1. *Let $\hat{\beta}_1$ be the ordinary least squares estimate of β_1 . Then if the conditions of Assumption 2 are satisfied, $\hat{\beta}_1 - \beta_1 \xrightarrow{p} 0$ and $\sqrt{ST}(\hat{\beta}_1 - \beta_1) \xrightarrow{d} N(0, M^{-1}\Omega M^{-1})$, where*

$$\begin{aligned}M &= M_{XX} - M_{XZ}M_{ZZ}^{-1}M'_{XZ}, \\ \Omega &= \lim_{T \rightarrow \infty} \text{E}[\frac{1}{T}X'_s\Gamma(\alpha)X_s] - M_{XZ}M_{ZZ}^{-1}(\lim_{T \rightarrow \infty} \text{E}[\frac{1}{T}Z'_s\Gamma(\alpha)X_s]) \\ &\quad - (\lim_{T \rightarrow \infty} \text{E}[\frac{1}{T}X'_s\Gamma(\alpha)Z_s])M_{ZZ}^{-1}M'_{XZ} + M_{XZ}M_{ZZ}^{-1}(\lim_{T \rightarrow \infty} [\frac{1}{T}Z'_s\Gamma(\alpha)Z_s])M_{ZZ}^{-1}M'_{XZ},\end{aligned}$$

with $M_{XX} = \lim_{T \rightarrow \infty} \text{E}[\frac{1}{T}X'_sX_s]$, $M_{XZ} = \lim_{T \rightarrow \infty} \text{E}[\frac{1}{T}X'_sZ_s]$, and $M_{ZZ} = \lim_{T \rightarrow \infty} [\frac{1}{T}Z'_sZ_s]$.

Proof. Let $Q_s = I_T - Z_s(Z'_sZ_s)^{-1}Z'_s$. Then

$$\text{E}\|\frac{1}{T}X'_sQ_sV_s\|^{1+\delta} \leq \text{E}(\|\frac{1}{\sqrt{T}}Q_sX_s\|^{1+\delta}\|\frac{1}{\sqrt{T}}Q_sV_s\|^{1+\delta}) \leq (\text{E}\|\frac{1}{\sqrt{T}}X_s\|^{2+2\delta}\text{E}\|\frac{1}{\sqrt{T}}V_s\|^{2+2\delta})^{1/2} \leq k_1^{\frac{1+\delta}{2}}\Delta,$$

where the first inequality is from $\|AB\| \leq \|A\|\|B\|$, the second from the Cauchy-Schwarz inequality and $\|Q_s A\| \leq \|A\|$, and the third from

$$\begin{aligned} \mathbb{E}\left\|\frac{1}{\sqrt{T}}X_s\right\|^{2+2\delta} &= \mathbb{E}\left[\frac{1}{T}\text{trace}(X'_s X_s)\right]^{1+\delta} \\ &= \frac{1}{T^{1+\delta}}\mathbb{E}\left[\sum_{t=1}^T \sum_{h=1}^{k_1} x_{tth}^2\right]^{1+\delta} \\ &\leq \frac{1}{T^{1+\delta}}\left[\sum_{t=1}^T \sum_{h=1}^{k_1} (\mathbb{E}|x_{tth}^2|^{1+\delta})^{\frac{1}{1+\delta}}\right]^{1+\delta} \\ &\leq \frac{1}{T^{1+\delta}}(Tk_1\Delta^{\frac{1}{1+\delta}})^{1+\delta} = k_1^{1+\delta}\Delta, \end{aligned}$$

where the first inequality follows from Minkowski's inequality and the second from NT5, and $\mathbb{E}\left\|\frac{1}{\sqrt{T}}V_s\right\|^{2+2\delta} \leq \Delta$ by a similar argument. It also follows that $\mathbb{E}\left\|\frac{1}{T}X'_s Q_s X_s\right\|^{1+\delta} \leq k_1^{1+\delta}\Delta$ by the same reasoning. So $\mathbb{E}\left\|\frac{1}{T}X'_s Q_s X_s\right\|$ and $\mathbb{E}\left\|\frac{1}{T}X'_s Q_s V_s\right\|$ are uniformly integrable in T . (See, e.g. Billingsley (1995).) Also, under NT3-NT5,

$$\begin{aligned} \frac{1}{T}\sum_{t=1}^T x_{st}x'_{st} - \frac{1}{T}\sum_{t=1}^T \mathbb{E}[x_{st}x'_{st}] &\xrightarrow{p} 0, \quad \frac{1}{T}\sum_{t=1}^T x_{st}z'_{st} - \frac{1}{T}\sum_{t=1}^T \mathbb{E}[x_{st}z'_{st}] \xrightarrow{p} 0 \\ \frac{1}{T}\sum_{t=1}^T z_{st}z'_{st} &\rightarrow \lim_{T \rightarrow \infty} \frac{1}{T}\sum_{t=1}^T z_{st}z'_{st}, \quad \frac{1}{T}\sum_{t=1}^T x_{st}v'_{st} \xrightarrow{p} 0, \quad \text{and} \quad \frac{1}{T}\sum_{t=1}^T z_{st}v'_{st} \xrightarrow{p} 0 \end{aligned}$$

by a LLN, e.g. White (2001) Corollary 3.48. It then follows from Phillips and Moon (1999) Corollary 1 that $\frac{1}{S}\sum_{s=1}^S \frac{1}{T}X'_s Q_s X_s \xrightarrow{p} M$ and $\frac{1}{S}\sum_{s=1}^S \frac{1}{T}X'_s Q_s V_s \xrightarrow{p} 0$. Hence, $\hat{\beta}_1 - \beta_1 \xrightarrow{p} 0$.

Let K denote a generic constant. To verify asymptotic normality, note that

$$\begin{aligned} \mathbb{E}\left\|\frac{1}{\sqrt{T}}X'_s V_s\right\|^{2+\frac{\delta}{2}} &\leq \frac{C}{T^{\frac{2+\frac{\delta}{2}}{2}}} \max\left\{\sum_{t=1}^T (\mathbb{E}|x_{st}v_{st}|^{2+\frac{\delta}{2}+\epsilon})^{\frac{2+\frac{\delta}{2}}{2+\frac{\delta}{2}+\epsilon}}, \left[\sum_{t=1}^T (\mathbb{E}|x_{st}v_{st}|^{2+\epsilon})^{\frac{2}{2+\epsilon}}\right]^{\frac{2+\frac{\delta}{2}}{2}}\right\} \\ &\leq CK < \infty \end{aligned} \tag{9}$$

by NT4-NT5 and Doukhan (1994) Theorem 2. Also,

$$\begin{aligned} \mathbb{E}\left\|\left(\frac{1}{T}X'_s Z_s\right)\left(\frac{1}{T}Z'_s Z_s\right)^{-1}\left(\frac{1}{\sqrt{T}}Z'_s V_s\right)\right\|^{2+\frac{\delta}{2}} &\leq (k_2 M)^{2+\frac{\delta}{2}} (\mathbb{E}\left\|\frac{1}{T}X'_s Z_s\right\|^{4+\delta} \mathbb{E}\left\|\frac{1}{\sqrt{T}}Z'_s V_s\right\|^{4+\delta})^{1/2} \\ &\leq CK < \infty, \end{aligned} \tag{10}$$

where the second inequality follows from $\|AB\| \leq \|A\|\|B\|$, the Cauchy-Schwarz inequality, and NT3, and the third inequality is from NT4-NT5 and Doukhan (1994) Theorem 2. (9) and (10) imply $\mathbb{E}\left\|\frac{1}{\sqrt{T}}X'_s Q_s V_s\right\|^{2+\frac{\delta}{2}} < \infty$ since

$$\mathbb{E}\left\|\frac{1}{\sqrt{T}}X'_s Q_s V_s\right\|^{2+\frac{\delta}{2}} \leq \left[\left(\mathbb{E}\left\|\frac{1}{\sqrt{T}}X'_s V_s\right\|^{2+\frac{\delta}{2}}\right)^{\frac{1}{2+\frac{\delta}{2}}} + \left(\mathbb{E}\left\|\frac{1}{\sqrt{T}}X'_s Z_s (Z'_s Z_s)^{-1} Z'_s V_s\right\|^{2+\frac{\delta}{2}}\right)^{\frac{1}{2+\frac{\delta}{2}}}\right]^{2+\frac{\delta}{2}}$$

by Minkowski's inequality which yields $\mathbb{E}\left\|\frac{1}{\sqrt{T}}X'_s Q_s V_s\right\|^2$ uniformly integrable in T . (See, e.g. Billingsley (1995).) $\frac{1}{\sqrt{T}}X'_s Q_s V_s \xrightarrow{d} N(0, \Omega)$ from White (2001) Theorem 5.20, from which it follows that $\Omega_T = \frac{1}{T}\mathbb{E}[X'_s Q_s \Gamma(\alpha) Q_s X_s] \rightarrow \Omega$. (E.g. Billingsley (1995) Corollary 25.12.) Then Phillips and

Moon (1999) Theorem 3 gives $\frac{1}{\sqrt{ST}} \sum_{s=1}^S \sum_{t=1}^T X'_s Q_s V_s \xrightarrow{d} N(0, \Omega)$ which implies $\sqrt{ST}(\hat{\beta}_1 - \beta_1) \xrightarrow{d} N(0, M^{-1}\Omega M^{-1})$. ■

Lemma C.2. Define \tilde{v}_{st} to be the residual from least squares regression of (1); i.e. $\tilde{v}_{st} = C_{st} - x'_{st}\hat{\beta}_1 - z'_{st}\hat{\beta}_2^s = v_{st} - x'_{st}(\hat{\beta}_1 - \beta_1) - z'_{st}(\hat{\beta}_2^s - \beta_2) = \ddot{v}_{st} - \ddot{x}'_{st}(\hat{\beta}_1 - \beta_1)$, where $\hat{\beta}_1$ and $\hat{\beta}_2^s$ are least squares estimates of β_1 and β_2^s . Let \tilde{v}_{st}^- be a $p \times 1$ vector with $\tilde{v}_{st}^- = [\tilde{v}_{s(t-p)}, \dots, \tilde{v}_{s(t-1)}]'$, and let \ddot{v}_{st}^- be a $p \times 1$ vector with $\ddot{v}_{st}^- = [\ddot{v}_{s(t-p)}, \dots, \ddot{v}_{s(t-1)}]'$. Under the conditions of Assumption 2,

$$\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \tilde{v}_{st}^- \tilde{v}_{st}^{-\prime} = \frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-\prime} + o_p((ST)^{-1/2}),$$

and

$$\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \tilde{v}_{st}^- \tilde{v}_{st} = \frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st} + o_p((ST)^{-1/2}).$$

Proof. Using calculations similar to those found in the proof of Lemma C.1, it can be shown that $\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st} \ddot{x}_{s(t-j)h} = o_p(1)$ and $\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{x}_{st} \ddot{x}_{s(t-j)h} = O_p(1)$ for $j = -p, \dots, p$ and $h = 1, \dots, k_1$. The conclusion then follows by Lemma C.1 and the same arguments used in proving Lemma B.2. ■

Lemma C.3. Under the conditions of Assumption 2,

- (i) $\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-\prime} \xrightarrow{p} \Gamma_p(\alpha)$, and $\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st} \xrightarrow{p} A(\alpha)$.
- (ii) $\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-\prime} = \frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T v_{st}^- v_{st}^{-\prime} + O_p(\frac{1}{T})$, and $\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st} = \frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T v_{st}^- v_{st} + O_p(\frac{1}{T})$.

Proof. The proof is given for $\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}^{-\prime} \xrightarrow{p} \Gamma_p(\alpha)$; $\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st} \xrightarrow{p} A(\alpha)$ follows by a similar argument.

$\frac{1}{T-p} \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}'$ is a $p \times p$ matrix with i, j element

$$\begin{aligned}
\left[\frac{1}{T-p} \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}' \right]_{[i,j]} &= \frac{1}{T-p} \sum_{t=p+1}^T v_{s(t-i)} v_{s(t-j)} \\
&- \frac{1}{T-p} \sum_{t=p+1}^T v_{s(t-i)} z'_{s(t-j)} (Z'_s Z_s)^{-1} Z'_s V_s \\
&- \frac{1}{T-p} \sum_{t=p+1}^T v_{s(t-j)} z'_{s(t-i)} (Z'_s Z_s)^{-1} Z'_s V_s \\
&+ \frac{1}{T-p} \sum_{t=p+1}^T z'_{s(t-i)} (Z'_s Z_s)^{-1} Z'_s V_s V'_s Z'_s (Z'_s Z_s)^{-1} z_{s(t-j)} \\
&= \frac{1}{T-p} \sum_{t=p+1}^T v_{s(t-i)} v_{s(t-j)} \\
&- \frac{1}{T-p} V'_{s,-i} Z_{s,-j} (Z'_s Z_s)^{-1} Z'_s V_s \\
&- \frac{1}{T-p} V'_{s,-j} Z_{s,-i} (Z'_s Z_s)^{-1} Z'_s V_s \\
&+ \frac{1}{T-p} Z_{s,-i} (Z'_s Z_s)^{-1} Z'_s V_s V'_s Z'_s (Z'_s Z_s)^{-1} Z'_{s,-j},
\end{aligned} \tag{11}$$

where $V_{s,-k} = (v_{s(p+1-k)}, v_{s(p+2-k)}, \dots, v_{s(T-k)})'$ and $Z_{s,-k}$ is defined similarly.

Consider

$$\begin{aligned}
&\mathbb{E} \| V'_s Z_s (Z'_s Z_s)^{-1} Z'_{s,-i} Z_{s,-j} (Z'_s Z_s)^{-1} Z'_s V_s \|^2 \\
&= \frac{T-p}{T} \mathbb{E} \left\| \left(\frac{1}{\sqrt{T}} V'_s Z_s \right) \left(\frac{1}{T} Z'_s Z_s \right)^{-1} \left(\frac{1}{T-p} Z'_{s,-i} Z_{s,-j} \right) \left(\frac{1}{T} Z'_s Z_s \right)^{-1} \left(\frac{1}{\sqrt{T}} Z'_s V_s \right) \right\|^2.
\end{aligned} \tag{12}$$

Note that

$$\left\| \frac{1}{T-p} Z'_{s,-i} Z_{s,-j} \right\| \leq k_2 \Delta < \infty \tag{13}$$

by NT5. Also,

$$\begin{aligned}
\mathbb{E} \left\| \left(\frac{1}{\sqrt{T}} Z'_s V_s \right) \right\|^{2+\delta} &= \frac{1}{T^{\frac{2+\delta}{2}}} \mathbb{E} \left\| \left(\sum_{t=1}^T z_{st} v_{st} \right) \right\|^{2+\delta} \\
&\leq \frac{1}{T^{\frac{2+\delta}{2}}} C \max \left\{ \sum_{t=1}^T (\mathbb{E} |z_{st} v_{st}|^{2+\delta+\epsilon})^{\frac{2+\delta}{2+\delta+\epsilon}}, \left[\sum_{t=1}^T (\mathbb{E} |z_{st} v_{st}|^{2+\epsilon})^{\frac{2}{2+\epsilon}} \right]^{\frac{2+\delta}{2}} \right\} \\
&\leq K < \infty
\end{aligned} \tag{14}$$

by Doukhan (1994) Theorem 2 and NT5. It then follows from the Holder's inequality, the Cauchy-Schwarz inequality, $\|AB\| \leq \|A\| \|B\|$, (13), (14), and NT3-NT5 that (12) is bounded away from ∞ , which gives $\|V'_s Z_s (Z'_s Z_s)^{-1} Z'_{s,-i} Z_{s,-j} (Z'_s Z_s)^{-1} Z'_s V_s\|$ uniformly integrable in T . Similarly, it can be

shown that $\|V'_{s,-i}Z_{s,-j}(Z'_sZ_s)^{-1}Z'_sV_s\|$ and $\|\frac{1}{T-p}\sum_{t=p+1}^Tv_{s(t-i)}v_{s(t-j)}\|$ are uniformly integrable in T .

Also, as $T \rightarrow \infty$,

1. $\frac{1}{T-p}\sum_{t=p+1}^Tv_{s(t-i)}v_{s(t-j)} \xrightarrow{p} \gamma_{|i-j|}(\alpha)$,
2. $V'_{s,-i}Z_{s,-j}(Z'_sZ_s)^{-1}Z'_sV_s \xrightarrow{d} \Psi'_{ij}M_{ZZ}^{-1}\Psi$,
3. $V'_sZ_s(Z'_sZ_s)^{-1}Z'_{s,-i}Z_{s,-j}(Z'_sZ_s)^{-1}Z'_sV_s \xrightarrow{d} \Psi'M_{ZZ}^{-1}M_{ij}M_{ZZ}^{-1}\Psi$,

where $M_{ZZ} = \lim_{T \rightarrow \infty} \frac{1}{T}[Z'_sZ_s]$, $M_{ij} = \lim_{T \rightarrow \infty} \frac{1}{T}[Z'_{s,-i}Z_{s,-j}]$, $\Psi \sim N(0, \lim_{T \rightarrow \infty} \frac{1}{T}[Z'_s\Gamma(\alpha)Z_s]$, $\Psi_{ij} \sim N(0, \lim_{T \rightarrow \infty} \frac{1}{T-p}[Z'_{s,-j}\Gamma(\alpha)Z_{s,-j}]$, $\Gamma_{-k} = E[v_s v'_{s,-k}]$, and $E[\Psi\Psi'_{ij}] = \lim_{T \rightarrow \infty} \frac{1}{T}[Z'_s\Gamma_{-i}(\alpha)Z_{s,-j}]$. (See White (2001) Corollary 3.48 and Theorem 5.20.) Combining 1., 2., and 3. and (11) gives

$$\frac{1}{T-p}\sum_{t=p+1}^T\ddot{v}_{s(t-i)}\ddot{v}_{s(t-j)} \xrightarrow{p} \gamma_{|i-j|}(\alpha).$$

It then follows from Phillips and Moon (1999) Corollary 1 that

$$\frac{1}{S(T-p)}\sum_{s=1}^S\sum_{t=p+1}^T\ddot{v}_{s(t-i)}\ddot{v}_{s(t-j)} \xrightarrow{p} \gamma_{|i-j|}(\alpha),$$

which yields the first conclusion. Also, using Phillips and Moon (1999) Corollary 1 and considering only 2., 3. and (11) yields the second conclusion. ■

Lemma C.4. Let $\hat{\alpha} = (\frac{1}{S(T-p)}\sum_{s=1}^S\sum_{t=p+1}^T\tilde{v}_{st}\tilde{v}'_{st})^{-1}(\frac{1}{S(T-p)}\sum_{s=1}^S\sum_{t=p+1}^T\tilde{v}_{st}\tilde{v}_{st})$ be the least squares estimate of α using the least squares residuals, \tilde{v}_{st} from estimating β_1 . If Assumption 2 is satisfied,

$$\begin{aligned}\hat{\alpha} &= \left(\frac{1}{S(T-p)}\sum_{s=1}^S\sum_{t=p+1}^T\ddot{v}_{st}\ddot{v}'_{st}\right)^{-1}\left(\frac{1}{S(T-p)}\sum_{s=1}^S\sum_{t=p+1}^T\ddot{v}_{st}\ddot{v}_{st}\right) + o_p((ST)^{-1/2}) \\ &= \left(\frac{1}{S(T-p)}\sum_{s=1}^S\sum_{t=p+1}^Tv_{st}v'_{st}\right)^{-1}\left(\frac{1}{S(T-p)}\sum_{s=1}^S\sum_{t=p+1}^Tv_{st}v_{st}\right) + O_p(T^{-1}) + o_p((ST)^{-1/2}).\end{aligned}$$

Proof. Follows immediately from Lemmas C.2 and C.3 using the same argument as in Lemma B.4. ■

Lemma C.5. For $\alpha_T(\alpha)$ defined in Lemma 1, $\alpha_T(\alpha) - \alpha = \frac{1}{T-p}B(\alpha, T)$, where $B(\alpha, T) \rightarrow B(\alpha)$ as $T \rightarrow \infty$, and $\alpha_T(\hat{\alpha}) - \hat{\alpha} = \frac{1}{T-p}B(\hat{\alpha}, T)$ if the conditions of Assumption 2 are met.

Proof. Recall $\alpha_T(\alpha) = (\Gamma_p(\alpha) + \frac{1}{T-p}\Delta\Gamma(\alpha))^{-1}(A(\alpha) + \frac{1}{T-p}\Delta A(\alpha))$ and that $\alpha = \Gamma_p(\alpha)^{-1}A(\alpha)$. Then $\alpha_T(\alpha) - \alpha = \frac{1}{T-p}(\Gamma_p(\alpha) + \frac{1}{T-p}\Delta\Gamma(\alpha))^{-1}(\Delta A(\alpha) + \Delta\Gamma(\alpha)\alpha) = \frac{1}{T-p}B(\alpha, T)$. That $B(\alpha, T) \rightarrow B(\alpha)$ as $T \rightarrow \infty$ follows from the proof of Lemma C.3. $\alpha_T(\hat{\alpha}) - \hat{\alpha} = \frac{1}{T-p}B(\hat{\alpha}, T)$ follows in the same fashion. ■

Lemma C.6. Define $\ddot{\mu}_{st} = \ddot{v}_{st} - \ddot{v}_{st}' \alpha_T(\alpha)$. If Assumption 2 is satisfied,

$$\frac{1}{\sqrt{S(T-p)}} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st} \ddot{\mu}_{st} = \frac{1}{\sqrt{S(T-p)}} \sum_{s=1}^S \sum_{t=p+1}^T v_{st} \eta_{st} + o_p(1).$$

Proof. $\frac{1}{\sqrt{S(T-p)}} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st} \ddot{\mu}_{st} = \frac{1}{\sqrt{S(T-p)}} \sum_{s=1}^S \sum_{t=p+1}^T (\ddot{v}_{st} \ddot{\mu}_{st} - \mathbb{E}[\ddot{v}_{st} \ddot{\mu}_{st}])$ since

$$\begin{aligned} \mathbb{E}\left[\sum_{t=p+1}^T \ddot{v}_{st} \ddot{\mu}_{st}\right] &= \mathbb{E}\left[\sum_{t=p+1}^T \ddot{v}_{st} (\ddot{v}_{st} - \ddot{v}_{st}' \alpha_T(\alpha))\right] \\ &= \mathbb{E}\left[\sum_{t=p+1}^T \ddot{v}_{st} \ddot{v}_{st}\right] - \mathbb{E}\left\{\sum_{t=p+1}^T \ddot{v}_{st} \ddot{v}_{st}' (\mathbb{E}\left[\sum_{t=p+1}^T \ddot{v}_{st} \ddot{v}_{st}'\right])^{-1} \mathbb{E}\left[\sum_{t=p+1}^T \ddot{v}_{st} \ddot{v}_{st}\right]\right\} \\ &= 0, \end{aligned}$$

where the second equality comes from $\alpha_T(\alpha) = \mathbb{E}[\sum_{t=p+1}^T \ddot{v}_{st} \ddot{v}_{st}']^{-1} \mathbb{E}[\sum_{t=p+1}^T \ddot{v}_{st} \ddot{v}_{st}]$ from Proposition 1. Also note that $\ddot{\mu}_{st} = \eta_{st} - (v_{st} - \ddot{v}_{st}) - v_{st}'(\alpha_T(\alpha) - \alpha) - (\ddot{v}_{st} - v_{st}')\alpha_T(\alpha)$ and $v_{st}^- = v_{st}^- - Z_{st}^-(Z_s' Z_s)^{-1} Z_s' V_s$ where $Z_{st}^- = [z_{s(t-1)}, \dots, z_{s(t-p)}]'$. So,

$$\begin{aligned} \ddot{v}_{st} \ddot{\mu}_{st} - \mathbb{E}[\ddot{v}_{st} \ddot{\mu}_{st}] &= (v_{st}^- - Z_{st}^-(Z_s' Z_s)^{-1} Z_s' V_s) \times \\ &\quad (\eta_{st} - z_{st}'(Z_s' Z_s)^{-1} Z_s' V_s - v_{st}'(\alpha_T(\alpha) - \alpha) + V_s' Z_s (Z_s' Z_s)^{-1} Z_{st}' \alpha_T(\alpha)) \\ &\quad - \mathbb{E}[(v_{st}^- - Z_{st}^-(Z_s' Z_s)^{-1} Z_s' V_s) \times \\ &\quad (\eta_{st} - z_{st}'(Z_s' Z_s)^{-1} Z_s' V_s - v_{st}'(\alpha_T(\alpha) - \alpha) + V_s' Z_s (Z_s' Z_s)^{-1} Z_{st}' \alpha_T(\alpha))] \\ &= v_{st}^- \eta_{st} - (v_{st}^- z_{st}'(Z_s' Z_s)^{-1} Z_s' V_s - \mathbb{E}[v_{st}^- z_{st}'(Z_s' Z_s)^{-1} Z_s' V_s]) \\ &\quad - (v_{st}^- v_{st}' - \mathbb{E}[v_{st}^- v_{st}']) (\alpha_T(\alpha) - \alpha) \\ &\quad + (v_{st}^- V_s' Z_s (Z_s' Z_s)^{-1} Z_{st}' - \mathbb{E}[v_{st}^- V_s' Z_s (Z_s' Z_s)^{-1} Z_{st}']) \alpha_T(\alpha) \\ &\quad - (Z_{st}^-(Z_s' Z_s)^{-1} Z_s' V_s \eta_{st} - \mathbb{E}[(Z_{st}^-(Z_s' Z_s)^{-1} Z_s' V_s \eta_{st})]) \\ &\quad + (Z_{st}^-(Z_s' Z_s)^{-1} Z_s' V_s z_{st}'(Z_s' Z_s)^{-1} Z_s' V_s \\ &\quad \quad - \mathbb{E}[Z_{st}^-(Z_s' Z_s)^{-1} Z_s' V_s z_{st}'(Z_s' Z_s)^{-1} Z_s' V_s]) \\ &\quad + (Z_{st}^-(Z_s' Z_s)^{-1} Z_s' V_s v_{st}' - \mathbb{E}[Z_{st}^-(Z_s' Z_s)^{-1} Z_s' V_s v_{st}']) (\alpha_T(\alpha) - \alpha) \\ &\quad - (Z_{st}^-(Z_s' Z_s)^{-1} Z_s' V_s V_s' Z_s (Z_s' Z_s)^{-1} Z_{st}' \\ &\quad \quad - \mathbb{E}[Z_{st}^-(Z_s' Z_s)^{-1} Z_s' V_s V_s' Z_s (Z_s' Z_s)^{-1} Z_{st}']) \alpha_T(\alpha). \end{aligned} \tag{15}$$

Now consider

$$\begin{aligned} Y_{S,T} &= \frac{1}{\sqrt{S(T-p)}} \sum_{s=1}^S \sum_{t=p+1}^T (Z_{st}^-(Z_s' Z_s)^{-1} Z_s' V_s V_s' Z_s (Z_s' Z_s)^{-1} Z_{st}' \\ &\quad - \mathbb{E}[Z_{st}^-(Z_s' Z_s)^{-1} Z_s' V_s V_s' Z_s (Z_s' Z_s)^{-1} Z_{st}']). \end{aligned}$$

$Y_{S,T}$ is a $p \times p$ matrix with i, j element

$$\begin{aligned} [Y_{S,T}]_{[i,j]} &= \frac{1}{\sqrt{S(T-p)}} \sum_{s=1}^S \sum_{t=p+1}^T (z'_{s(t-i)} (Z'_s Z_s)^{-1} Z'_s V_s V'_s Z_s (Z'_s Z_s)^{-1} z_{s(t-j)} \\ &\quad - \mathbb{E}[z_{s(t-i)} (Z'_s Z_s)^{-1} Z'_s V_s V'_s Z_s (Z'_s Z_s)^{-1} z_{s(t-j)}]) \\ &= \frac{1}{\sqrt{S(T-p)}} \sum_{s=1}^S (V'_s Z_s (Z'_s Z_s)^{-1} Z'_{s,-i} Z_{s,-j} (Z'_s Z_s)^{-1} Z'_s V_s \\ &\quad - \mathbb{E}[V'_s Z_s (Z'_s Z_s)^{-1} Z'_{s,-i} Z_{s,-j} (Z'_s Z_s)^{-1} Z'_s V_s]), \end{aligned}$$

and

$$\begin{aligned} \mathbb{E}\| [Y_{S,T}]_{[i,j]} \|^2 &= \frac{1}{T-p} \mathbb{E}\| V'_s Z_s (Z'_s Z_s)^{-1} Z'_{s,-i} Z_{s,-j} (Z'_s Z_s)^{-1} Z'_s V_s \|^2 \\ &\quad - \frac{1}{T-p} \|\mathbb{E}[V'_s Z_s (Z'_s Z_s)^{-1} Z'_{s,-i} Z_{s,-j} (Z'_s Z_s)^{-1} Z'_s V_s]\|^2 \rightarrow 0, \end{aligned}$$

where the equality is from independence assumption NT2, since $\mathbb{E}\|A\|^2 \geq \|E[A]\|^2$ and

1. $\mathbb{E}\| V'_s Z_s (Z'_s Z_s)^{-1} Z'_{s,-i} Z_{s,-j} (Z'_s Z_s)^{-1} Z'_s V_s \|^2$
 $\leq \frac{T-p}{T} (\mathbb{E}\| \frac{1}{\sqrt{T}} V'_s Z_s \|^4) (\| \frac{1}{T} Z'_s Z_s \|^4) (\| \frac{1}{T-p} Z'_{s,-i} Z_{s,-j} \|^4) (\| \frac{1}{T} Z'_s Z_s \|^4) \mathbb{E}\| \frac{1}{\sqrt{T}} Z'_s V_s \|^4)^{1/2}$
 by the Cauchy-Schwarz inequality.
2. $\| (\frac{1}{T} Z'_s Z_s)^{-1} (\frac{1}{T-p} Z'_{s,-i} Z_{s,-j}) (\frac{1}{T} Z'_s Z_s)^{-1} \|^2 \leq (k_2 M)^2 (\| \frac{1}{T-p} Z'_{s,-i} Z_{s,-j} \|^2) \leq (k_2 M)^2 k_2 \Delta$ by NT3 and NT5.
3. $\mathbb{E}\| \frac{1}{\sqrt{T}} Z'_s V_s \|^4 \leq T^{-2} C \max\{ \sum_{t=1}^T (\mathbb{E}|z_{st} v_{st}|^{4+\epsilon})^{\frac{4}{4+\epsilon}}, [\sum_{t=1}^T (\mathbb{E}|z_{st} v_{st}|^{2+\epsilon})^{\frac{2}{2+\epsilon}}]^2 \} \leq K$ by Doukhan (1994) Theorem 2 and NT5.

Hence, $[Y_{S,T}]_{[i,j]} \xrightarrow{P} 0$ by Chebychev's inequality, and it follows that $Y_{S,T} = o_p(1)$. That all other terms, except $\frac{1}{\sqrt{S(T-p)}} \sum_{s=1}^S \sum_{t=p+1}^T (v_{st}^- v_{st}^{-'} - \mathbb{E}[v_{st}^- v_{st}^{-'}]) (\alpha_T(\alpha) - \alpha)$, are $o_p(1)$ follows similarly.

To show $\frac{1}{\sqrt{S(T-p)}} \sum_{s=1}^S \sum_{t=p+1}^T (v_{st}^- v_{st}^{-'} - \mathbb{E}[v_{st}^- v_{st}^{-'}]) (\alpha_T(\alpha) - \alpha) = o_p(1)$, note that $(\alpha_T(\alpha) - \alpha) = O(\frac{1}{T})$ from Lemma C.5. Also, $\text{vec}\{ \frac{1}{\sqrt{T-p}} \sum_{t=p+1}^T (v_{st}^- v_{st}^{-'} - \mathbb{E}[v_{st}^- v_{st}^{-'}]) \} \xrightarrow{d} \Psi \sim N(0, \Omega)$ where

$$\Omega = \lim_{T \rightarrow \infty} \frac{1}{T-p} \sum_{t_1=p+1}^T \sum_{t_2=p+1}^T [\mathbb{E}(v_{st_1}^- v_{st_2}^{-'} \otimes v_{st_1}^- v_{st_2}^{-'}) - \mathbb{E}(v_{st_1}^- \otimes v_{st_1}^-) \mathbb{E}(v_{st_2}^{-'} \otimes v_{st_2}^{-'})]$$

as $T \rightarrow \infty$ follows from a CLT (e.g. White (2001) Theorem 5.20) and the Cramer-Wold device, and $\| \frac{1}{\sqrt{T-p}} \sum_{t=p+1}^T \text{vec}(v_{st}^- v_{st}^{-'} - \mathbb{E}[v_{st}^- v_{st}^{-'}]) \|^2 \xrightarrow{d} \|\Psi\|^2$ by the Continuous Mapping Theorem. In

addition,

$$\begin{aligned}
& \mathbb{E} \left\| \frac{1}{\sqrt{S(T-p)}} \sum_{s=1}^S \sum_{t=p+1}^T (v_{st}^- v_{st}' - \mathbb{E}[v_{st}^- v_{st}']) \right\|^2 \\
&= \text{trace} \left\{ \mathbb{E} \left[\frac{1}{T-p} \sum_{t_1=p+1}^T \sum_{t_2=p+1}^T ((v_{st_1}^- \otimes v_{st_1}^-) - \mathbb{E}(v_{st_1}^- \otimes v_{st_1}^-)) ((v_{st_2}^- \otimes v_{st_2}^-) - \mathbb{E}(v_{st_2}^- \otimes v_{st_2}^-))' \right] \right\} \\
&= \frac{1}{T-p} \sum_{t_1=p+1}^T \sum_{t_2=p+1}^T [\mathbb{E}(v_{st_1}^- v_{st_2}' \otimes v_{st_1}^- v_{st_2}') - \mathbb{E}(v_{st_1}^- \otimes v_{st_1}^-) \mathbb{E}(v_{st_2}' \otimes v_{st_2}')] \rightarrow \text{trace}(\Omega) = \mathbb{E} \|\Psi\|^2
\end{aligned}$$

as $T \rightarrow \infty$. It follows that $\left\| \frac{1}{\sqrt{S(T-p)}} \sum_{s=1}^S \sum_{t=p+1}^T (v_{st}^- v_{st}' - \mathbb{E}[v_{st}^- v_{st}']) \right\|^2$ is uniformly integrable in T . (For example, Billingsley (1995) 16.14.) Then, using Phillips and Moon (1999) Theorem 3,

$$\frac{1}{\sqrt{S(T-p)}} \sum_{s=1}^S \sum_{t=p+1}^T (v_{st}^- v_{st}' - \mathbb{E}[v_{st}^- v_{st}']) \xrightarrow{d} N(0, \Omega) \Rightarrow \frac{1}{\sqrt{S(T-p)}} \sum_{s=1}^S \sum_{t=p+1}^T (v_{st}^- v_{st}' - \mathbb{E}[v_{st}^- v_{st}']) = O_p(1),$$

and the conclusion follows immediately. ■

Lemma C.7. *If Assumption 2 is satisfied, $\frac{1}{\sqrt{S(T-p)}} \sum_{s=1}^S \sum_{t=p+1}^T v_{st}^- \eta_{st} \xrightarrow{d} N(0, \Xi)$, where*

$$\Xi = \lim_{T \rightarrow \infty} \frac{1}{T-p} \sum_{t_1=p+1}^T \sum_{t_2=p+1}^T \mathbb{E}[v_{st_1}^- \eta_{st_1} \eta_{st_2}' v_{st_2}'].$$

In addition, if η_{st} are independent for all s and t , $\Xi = \sigma_\eta^2 \Gamma$.

Proof. As $T \rightarrow \infty$, $\frac{1}{\sqrt{T-p}} \sum_{t=p+1}^T v_{st}^- \eta_{st} \xrightarrow{d} \Psi \sim N(0, \Xi)$ by a CLT (e.g. White (2001) Theorem 5.20) and the Cramer-Wold device, and $\left\| \frac{1}{\sqrt{T-p}} \sum_{t=p+1}^T v_{st}^- \eta_{st} \right\|^2 \xrightarrow{d} \|\Psi\|^2$ by the Continuous Mapping Theorem. Also, $\mathbb{E} \left\| \frac{1}{\sqrt{T-p}} \sum_{t=p+1}^T v_{st}^- \eta_{st} \right\|^2 = \text{trace} \left(\frac{1}{T-p} \sum_{t_1=p+1}^T \sum_{t_2=p+1}^T \mathbb{E} v_{st_1}^- \eta_{st_1} \eta_{st_2}' v_{st_2}^- \right) \rightarrow \text{trace}(\Xi) = \mathbb{E} \|\Psi\|^2$, which implies that $\mathbb{E} \left\| \frac{1}{\sqrt{T-p}} \sum_{t=p+1}^T v_{st}^- \eta_{st} \right\|^2$ is uniformly integrable in T . (For example, Billingsley (1995) 16.14.) Then, using Phillips and Moon (1999) Theorem 3,

$$\frac{1}{\sqrt{S(T-p)}} \sum_{s=1}^S \sum_{t=p+1}^T v_{st}^- \eta_{st} \xrightarrow{d} N(0, \Xi).$$

Also, if the η_{st} are independent for all s and t , $\mathbb{E} v_{st_1}^- \eta_{st_1} \eta_{st_2}' v_{st_2}^- = 0 \forall t_1 \neq t_2$, and $\mathbb{E} v_{st}^- \eta_{st}^2 v_{st}^- = \sigma_\eta^2 \Gamma$. ■

Lemma C.8. $\sqrt{ST}(\hat{\alpha} - \alpha_T(\alpha)) \xrightarrow{d} N(0, \Gamma^{-1} \Xi \Gamma^{-1})$ for Ξ defined in Lemma C.7 if Assumption 2 is satisfied.

Proof. From Lemma C.4, $\hat{\alpha} - \alpha_T(\alpha) = \left(\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}' \right)^{-1} \left(\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st} \right) - \alpha_T(\alpha) + o_p((ST)^{-1/2}) = \left(\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}' \right)^{-1} \left(\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{\mu}_{st} \right) + o_p((ST)^{-1/2})$. The conclusion then follows immediately from Lemmas C.3, C.6, and C.7. ■

Lemma C.9. *Under Assumption 2, $\hat{\alpha} - \alpha = \max\{O_p(T^{-1}), O_p((ST)^{-1/2})\}$.*

Proof. Lemma C.4 gives

$$\begin{aligned}\hat{\alpha} &= \left(\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T v_{st}^- v_{st}^{-'}\right)^{-1} \left(\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T v_{st}^- v_{st}\right) + O_p(T^{-1}) + o_p((ST)^{-1/2}) \\ &= \left(\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T v_{st}^- v_{st}^{-'}\right)^{-1} \left(\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T v_{st}^- \eta_{st}\right) + \alpha + O_p(T^{-1}) + o_p((ST)^{-1/2}).\end{aligned}$$

$\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T v_{st}^- \eta_{st} = O_p((ST)^{-1/2})$ from Lemma C.8, and $\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T v_{st}^- v_{st}^{-'} = O_p(1)$ follows immediately from a LLN, e.g. White (2001) Corollary 3.48. ■

APPENDIX D. PROOF OF PROPOSITION 6

The proof of Proposition 6 is quite similar to the proof provided in Rothenberg (1984) and is sketched below. Throughout, let

$$M_z(\tilde{\alpha}) = (I \otimes \Gamma(\tilde{\alpha})^{-1}) - (I \otimes \Gamma(\tilde{\alpha})^{-1})Z(Z'(I \otimes \Gamma(\tilde{\alpha})^{-1})Z)^{-1}Z'(I \otimes \Gamma(\tilde{\alpha})^{-1}).$$

Denote the GLS estimator of β_1 as

$$\hat{\beta}_1(\alpha) = (X'M_z(\alpha)X)^{-1}X'M_z(\alpha)C$$

and the FGLS estimator of β_1 corresponding to $\tilde{\alpha}$ as

$$\hat{\beta}_1(\tilde{\alpha}) = (X'M_z(\tilde{\alpha})X)^{-1}X'M_z(\tilde{\alpha})C.$$

Also, define

$$A(\tilde{\alpha}) = \frac{1}{ST}X'M_z(\tilde{\alpha})X,$$

and

$$\Psi(\tilde{\alpha}) = \frac{1}{\sqrt{ST}}X'M_z(\tilde{\alpha})MV$$

for $M = I - X(X'M_z(\alpha)X)^{-1}X'M_z(\alpha)$. Then for $i = 1, \dots, p$, $j = 1, \dots, p$, and $k = 1, \dots, p$, let

$$\begin{aligned}A_i &= \frac{\partial A}{\partial \tilde{\alpha}_i} \Big|_{\alpha}, & A_{ij} &= \frac{\partial^2 A}{\partial \tilde{\alpha}_i \partial \tilde{\alpha}_j} \Big|_{\alpha}, & A_{ijk} &= \frac{\partial^3 A}{\partial \tilde{\alpha}_i \partial \tilde{\alpha}_j \partial \tilde{\alpha}_k} \Big|_{\alpha}, \\ \Psi_i &= \frac{\partial \Psi}{\partial \tilde{\alpha}_i} \Big|_{\alpha}, & \Psi_{ij} &= \frac{\partial^2 \Psi}{\partial \tilde{\alpha}_i \partial \tilde{\alpha}_j} \Big|_{\alpha}, & \Psi_{ijk} &= \frac{\partial^3 \Psi}{\partial \tilde{\alpha}_i \partial \tilde{\alpha}_j \partial \tilde{\alpha}_k} \Big|_{\alpha}.\end{aligned}$$

Step 1. As in Rothenberg (1984), note that the residual vector

$$\tilde{V} = C - [X, Z]([X, Z]'[X, Z])^{-1}[X, Z]'C = V - [X, Z]([X, Z]'[X, Z])^{-1}[X, Z]'V$$

does not depend on β_1 and that $\hat{\alpha}$ and $\hat{\alpha}^{(\infty)}$ are even functions of \tilde{V} and hence V . For a given α , $\hat{\beta}_1(\alpha)$ is a complete sufficient statistic for β_1 since V is normally distributed. Also, by Basu's Theorem (See, for example, Lehmann (1983) p. 46), any statistic whose distribution does not depend on β_1 must be distributed independently of $\hat{\beta}_1(\alpha)$, so $\hat{\beta}_1(\alpha)$ is independent of $\hat{\alpha}$, $\hat{\alpha}^{(\infty)}$, $\hat{\beta}_1(\hat{\alpha}) - \hat{\beta}_1(\alpha)$, and $\hat{\beta}_1(\hat{\alpha}^{(\infty)}) - \hat{\beta}_1(\alpha)$. It then follows, for $\tilde{\alpha}$ equal to either $\hat{\alpha}$ or $\hat{\alpha}^{(\infty)}$, that

$$\mathbb{E}[(\hat{\beta}_1(\tilde{\alpha}) - \beta_1)(\hat{\beta}_1(\tilde{\alpha}) - \beta_1)'] = \mathbb{E}[(\hat{\beta}_1(\alpha) - \beta_1)(\hat{\beta}_1(\alpha) - \beta_1)'] + \mathbb{E}[(\hat{\beta}_1(\tilde{\alpha}) - \hat{\beta}_1(\alpha))(\hat{\beta}_1(\tilde{\alpha}) - \hat{\beta}_1(\alpha))'].$$

$\widehat{\beta}_1(\alpha) - \beta_1$ is exactly normal with variance $\frac{1}{ST}A^{-1}$, so the higher-order variance of $\widehat{\beta}_1(\widetilde{\alpha})$ is $\frac{1}{ST}A^{-1}$ plus the higher-order variance of $\widehat{\beta}_1(\widetilde{\alpha}) - \widehat{\beta}_1(\alpha)$. In addition, the bias of $\widehat{\beta}_1(\widetilde{\alpha})$ is

$$\mathbb{E}[\widehat{\beta}_1(\widetilde{\alpha}) - \beta_1] = \mathbb{E}[\widehat{\beta}_1(\alpha) - \beta_1] + \mathbb{E}[\widehat{\beta}_1(\widetilde{\alpha}) - \widehat{\beta}_1(\alpha)] = 0 + \mathbb{E}[\widehat{\beta}_1(\widetilde{\alpha}) - \widehat{\beta}_1(\alpha)],$$

so the higher-order bias is also determined by the higher-order bias of $\widehat{\beta}_1(\widetilde{\alpha}) - \widehat{\beta}_1(\alpha)$.

Step 2. Expansion of $\widehat{\beta}_1(\widetilde{\alpha}) - \widehat{\beta}_1(\alpha)$. For $\widetilde{\alpha}$ equal to either $\widehat{\alpha}$ or $\widehat{\alpha}^{(\infty)}$,

$$\widehat{\beta}_1(\widetilde{\alpha}) - \widehat{\beta}_1(\alpha) = A(\widetilde{\alpha})^{-1}\Psi(\widetilde{\alpha}). \quad (16)$$

Let $d_i = \sqrt{ST}(\widetilde{\alpha} - \alpha)_i$ be the i^{th} element of the vector $\sqrt{ST}(\widetilde{\alpha} - \alpha)$. Then expanding (16) about $\widetilde{\alpha} = \alpha$ and noting $\Psi(\alpha) = 0$ yields

$$\begin{aligned} \sqrt{ST}(\widehat{\beta}_1(\widetilde{\alpha}) - \widehat{\beta}_1(\alpha)) &= \sum_{i=1}^p A(\alpha)^{-1}\Psi_i(\alpha)d_i/\sqrt{ST} \\ &+ \frac{1}{2} \sum_{i=1}^p \sum_{j=1}^p (A(\alpha)^{-1}\Psi_{ij}(\alpha)d_id_j - A(\alpha)^{-1}A_i(\alpha)A(\alpha)^{-1}\Psi_j(\alpha)d_id_j \\ &- A(\alpha)^{-1}A_j(\alpha)A(\alpha)^{-1}\Psi_i(\alpha)d_id_j)/ST + R(\bar{\alpha})/(ST)^{3/2}, \end{aligned} \quad (17)$$

where $\bar{\alpha}$ is an intermediate value between α and $\widetilde{\alpha}$ and

$$\begin{aligned} R(\bar{\alpha}) &= \frac{1}{6} \sum_{i=1}^p \sum_{j=1}^p \sum_{k=1}^p [-6A(\bar{\alpha})^{-1}A_i(\bar{\alpha})A(\bar{\alpha})^{-1}A_j(\bar{\alpha})A(\bar{\alpha})^{-1}A_k(\bar{\alpha})A(\bar{\alpha})^{-1}\Psi(\bar{\alpha}) \\ &+ 3A(\bar{\alpha})^{-1}A_{ij}(\bar{\alpha})A(\bar{\alpha})^{-1}A_k(\bar{\alpha})A(\bar{\alpha})^{-1}\Psi(\bar{\alpha}) + 3A(\bar{\alpha})^{-1}A_i(\bar{\alpha})A(\bar{\alpha})^{-1}A_{jk}(\bar{\alpha})A(\bar{\alpha})^{-1}\Psi(\bar{\alpha}) \\ &+ 6A(\bar{\alpha})^{-1}A_i(\bar{\alpha})A(\bar{\alpha})^{-1}A_j(\bar{\alpha})A(\bar{\alpha})^{-1}\Psi_k(\bar{\alpha}) - A(\bar{\alpha})^{-1}A_{ijk}(\bar{\alpha})A(\bar{\alpha})^{-1}\Psi(\bar{\alpha}) \\ &- 3A(\bar{\alpha})^{-1}A_{ij}(\bar{\alpha})A(\bar{\alpha})^{-1}\Psi_k(\bar{\alpha}) - 3A(\bar{\alpha})^{-1}A_i(\bar{\alpha})A(\bar{\alpha})^{-1}\Psi_{jk}(\bar{\alpha}) \\ &+ A(\bar{\alpha})^{-1}\Psi_{ijk}(\bar{\alpha})]d_id_jd_k. \end{aligned} \quad (18)$$

Step 3. Expansion of $\widehat{\alpha} - \alpha_T(\alpha)$. Recall that the least squares estimator of α using the residuals from the estimation of equation (1), \widetilde{v}_{st} , is

$$\widehat{\alpha} = \left(\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \widetilde{v}_{st}\widetilde{v}_{st}' \right)^{-1} \left(\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \widetilde{v}_{st}\widetilde{v}_{st} \right) \equiv \widehat{G}_1^{-1}\widehat{G}_2$$

where $\widetilde{v}_{st}' = (\widetilde{v}_{s(t-p)}, \dots, \widetilde{v}_{s(t-1)})$ and $\widetilde{v}_{st} = \ddot{v}_{st} - \ddot{x}_{st}'(\widehat{\beta}_1 - \beta_1)$ for \ddot{v}_{st} and \ddot{x}_{st} as defined in (??) and (??) in the text and $\widehat{\beta}_1$ the least squares estimate of β_1 . Then

$$\begin{aligned} \widehat{G}_1 &= \frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}\ddot{v}_{st}' \\ &- \frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T (\ddot{v}_{st}^-(\widehat{\beta}_1 - \beta_1)' \ddot{x}_{st}^- \ddot{x}_{st}^-(\widehat{\beta}_1 - \beta_1) \ddot{v}_{st}^- + \ddot{x}_{st}^-(\widehat{\beta}_1 - \beta_1)(\widehat{\beta}_1 - \beta_1)' \ddot{x}_{st}^-) \\ &= \widehat{D}_1 + \widehat{M}_1 \end{aligned}$$

where $\widehat{D}_1 = \frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st}'$, and

$$\begin{aligned} \widehat{G}_2 &= \frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{v}_{st} \\ &\quad - \frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T (\ddot{v}_{st}^- (\widehat{\beta}_1 - \beta_1)' \ddot{x}_{st} \ddot{x}_{st}' (\widehat{\beta}_1 - \beta_1) \ddot{v}_{st} + \ddot{x}_{st}' (\widehat{\beta}_1 - \beta_1) (\widehat{\beta}_1 - \beta_1)' \ddot{x}_{st}) \\ &= \widehat{D}_2 + \widehat{D}_1 \alpha_T(\alpha) + \widehat{M}_2 \end{aligned}$$

where $\widehat{D}_2 = \frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \ddot{\mu}_{st}$ and $\ddot{\mu}_{st} = \ddot{v}_{st} - \ddot{v}_{st}' \alpha_T(\alpha)$. It then follows that

$$\begin{aligned} \widehat{\alpha} - \alpha_T(\alpha) &= \widehat{G}_1^{-1} \widehat{G}_2 - \alpha_T(\alpha) \\ &= (\widehat{D}_1 + \widehat{M}_1)^{-1} (\widehat{D}_2 + \widehat{M}_2 - \widehat{M}_1 \alpha_T(\alpha)). \end{aligned}$$

It is shown below that \widehat{M}_1 and \widehat{M}_2 are $O_p(1/ST)$, and it then follows from Lemmas C.3-C.9 that $\widehat{\alpha}$ and $\widehat{\alpha}^{(\infty)}$ are consistent and have the asymptotic distributions described in Proposition 3.

A Taylor series expansion then yields

$$\sqrt{ST}(\widehat{\alpha} - \alpha_T(\alpha)) = \Gamma_p^{-1} \sqrt{ST} \widehat{D}_2 \tag{19}$$

$$+ \Gamma_p^{-1} (\sqrt{ST} \widehat{M}_2 - \sqrt{ST} \widehat{M}_1 \alpha_T(\alpha)) - \Gamma_p^{-1} (\widehat{D}_1 - \Gamma_p) \Gamma_p^{-1} \sqrt{ST} \widehat{D}_2 \tag{20}$$

$$\begin{aligned} &- \Gamma_p^{-1} \widehat{M}_1 \Gamma_p^{-1} \sqrt{ST} \widehat{D}_2 - \Gamma_p^{-1} (\widehat{D}_1 - \Gamma_p) \Gamma_p^{-1} (\sqrt{ST} \widehat{M}_2 - \sqrt{ST} \widehat{M}_1 \alpha_T(\alpha)) \\ &+ \frac{1}{2} \Gamma_p^{-1} (\widehat{D}_1 - \Gamma_p) \Gamma_p^{-1} (\widehat{D}_1 - \Gamma_p) \Gamma_p^{-1} \sqrt{ST} \widehat{D}_2 \end{aligned} \tag{21}$$

$$\begin{aligned} &- \Gamma_p^{-1} \widehat{M}_1 \Gamma_p^{-1} (\sqrt{ST} \widehat{M}_2 - \sqrt{ST} \widehat{M}_1 \alpha_T(\alpha)) \\ &+ \frac{1}{2} \Gamma_p^{-1} \widehat{M}_1 \Gamma_p^{-1} (\widehat{D}_1 - \Gamma_p) \Gamma_p^{-1} \sqrt{ST} \widehat{D}_2 \\ &+ \frac{1}{2} \Gamma_p^{-1} (\widehat{D}_1 - \Gamma_p) \Gamma_p^{-1} \widehat{M}_1 \Gamma_p^{-1} \sqrt{ST} \widehat{D}_2 \\ &+ \frac{1}{2} \Gamma_p^{-1} \widehat{M}_1 \Gamma_p^{-1} \widehat{M}_1 \Gamma_p^{-1} \sqrt{ST} \widehat{D}_2 \\ &- \frac{1}{6} \widetilde{\Gamma}_p^{-1} (\widehat{D}_1 - \Gamma_p + \widehat{M}_1) \widetilde{\Gamma}_p^{-1} (\widehat{D}_1 - \Gamma_p + \widehat{M}_1) \widetilde{\Gamma}_p^{-1} (\widehat{D}_1 - \Gamma_p + \widehat{M}_1) \widetilde{\Gamma}_p^{-1} \times \\ &\quad (\sqrt{ST} \widehat{D}_2 + \sqrt{ST} \widehat{M}_2 - \sqrt{ST} \widehat{M}_1 \alpha_T(\alpha)) \end{aligned} \tag{22}$$

$$\equiv \psi_1 + \psi_2 / \sqrt{ST} + \psi_3 / ST + R / (ST)^{3/2}, \tag{23}$$

where $\widetilde{\Gamma}_p$ is a matrix of intermediate values that satisfy $\|\widetilde{\Gamma}_p - \Gamma_p\| \leq \|\widehat{D}_1 - \Gamma_p + \widehat{M}_1\|$.

It is shown in Lemmas C.3 and C.6 that $\widehat{D}_1 - \Gamma_p + \widehat{M}_1 \xrightarrow{p} 0$ and that $\widehat{D}_1 - \Gamma_p = O_p(1/\sqrt{ST})$, and it follows from Lemmas C.6 and C.7 that $\widehat{D}_2 = O_p(1/\sqrt{ST})$.

To show $\widehat{M}_2 = O_p(1/ST)$, let $M_Z = I - Z(Z'Z)^{-1}Z'$ and define M_{Z_s} similarly. Recall that

$$\begin{aligned}\widehat{M}_2 &= -\frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \dot{V}' \ddot{X} (\ddot{X}' \ddot{X})^{-1} \ddot{x}_{st} - \frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{x}_{st}' (\ddot{X}' \ddot{X})^{-1} \ddot{X}' \dot{V} \ddot{v}_{st} \\ &\quad + \frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{x}_{st}' (\ddot{X}' \ddot{X})^{-1} \ddot{X}' \dot{V} \dot{V}' \ddot{X} (\ddot{X}' \ddot{X})^{-1} \ddot{x}_{st}.\end{aligned}\tag{24}$$

Then

$$\begin{aligned}\mathbb{E} \|\dot{V}' \ddot{X} (\ddot{X}' \ddot{X})^{-1}\|^2 &\leq (\mathbb{E} \|\frac{1}{ST} V' M_Z X\|^2) \|\frac{1}{ST} (X' M_Z X)^{-1}\|^2 \\ &\leq \left(\frac{k_1^{1/2}}{\lambda_{\min}(\frac{1}{ST} X' M_Z X)} \right)^2 \frac{1}{(ST)^2} \sum_{s=1}^S \mathbb{E} \|V_s' M_{Z_s} X_s\| \\ &= \left(\frac{k_1^{1/2}}{\lambda_{\min}(\frac{1}{ST} X' M_Z X)} \right)^2 \frac{1}{(ST)^2} \sum_{s=1}^S \text{trace}(X_s' M_{Z_s} \Omega M_{Z_s} X_s) \\ &\leq \left(\frac{k_1^{1/2}}{\lambda_{\min}(\frac{1}{ST} X' M_Z X)} \right)^2 \frac{\lambda_{\max}(\Omega)}{(ST)^2} \sum_{s=1}^S \text{trace}(\Delta^2 i_{k_1} i_T' i_T i_{k_1}') \\ &= \left(\frac{k_1^{1/2}}{\lambda_{\min}(\frac{1}{ST} X' M_Z X)} \right)^2 \lambda_{\max}(\Omega) k_1 \Delta^2 / ST,\end{aligned}$$

where the first inequality follows from the definition of $\|A\|$ and the second inequality follows from boundedness of x_{sth} and $c'Ac \leq \lambda_{\max}(A)c'c$. Also, defining e_t to be a $T \times 1$ unit vector with t^{th} element equal to 1 and all other elements equal to 0 and $e_t^- = [e_{t-1}, \dots, e_{t-p}]'$,

$$\begin{aligned}\mathbb{E} \left\| \frac{1}{S(T-p)} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{x}'_{st} \otimes \ddot{v}_{st}^- \right\|^2 &= \frac{1}{(S(T-p))^2} \sum_{s=1}^S \sum_{t=p+1}^T \sum_{\tau=p+1}^T \text{trace}(\mathbb{E}(\ddot{x}'_{st} \ddot{x}_{s\tau} \otimes \ddot{v}_{st}^- \ddot{v}_{s\tau}^-)) \\ &\leq \frac{k_1 M^2}{(S(T-p))^2} \sum_{s=1}^S \text{trace}((\sum_{t=p+1}^T e_t^-) M_{Z_s} \Omega M_{Z_s} (\sum_{t=p+1}^T e_t^-')) \\ &\leq \frac{k_1 M^2 \lambda_{\max}(\Omega)}{(S(T-p))^2} \sum_{s=1}^S \text{trace}((\sum_{t=p+1}^T e_t^-) (\sum_{t=p+1}^T e_t^-')) \\ &\leq \frac{k_1 p M^2 \lambda_{\max}(\Omega)}{S(T-p)},\end{aligned}$$

where the first inequality results from boundedness of x_{sth} , the second from $c'Ac \leq \lambda_{\max}(A)c'c$, and the third from the definition of e_t^- . It then follows that

$$\begin{aligned}\frac{1}{ST} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{v}_{st}^- \dot{V}' \ddot{X} (\ddot{X}' \ddot{X})^{-1} \ddot{x}_{st} &= \left(\frac{1}{ST} \sum_{s=1}^S \sum_{t=p+1}^T \ddot{x}'_{st} \otimes \ddot{v}_{st}^- \right) \text{vec}(V' \ddot{X} (\ddot{X}' \ddot{X})^{-1}) \\ &= O_p(1/\sqrt{ST}) O_p(1/\sqrt{ST}) = O_p(1/ST).\end{aligned}$$

Using similar arguments, it can be shown that the second and third terms of expression (24) are $O_p(1/ST)$. It then follows that $\widehat{M}_2 = O_p(1/ST)$. That $\widehat{M}_1 = O_p(1/ST)$ follows similarly.

Finally, making use of $\widehat{M}_1 = O_p(1/ST)$, $\widehat{M}_2 = O_p(1/ST)$, $\widehat{D}_1 - \Gamma_p = O_p(1/\sqrt{ST})$, and $\widehat{D}_2 = O_p(1/\sqrt{ST})$ yields that ψ_1, ψ_2, ψ_3 , and R are all $O_p(1)$.

Step 4. Expansion of $\widehat{\alpha} - \alpha$ and $\widehat{\alpha}^{(\infty)} - \alpha$.

i. $\sqrt{ST}(\widehat{\alpha} - \alpha) = \sqrt{ST}(\widehat{\alpha} - \alpha_T(\alpha)) + \sqrt{ST}(\alpha_T(\alpha) - \alpha) = \sqrt{ST}(\widehat{\alpha} - \alpha_T(\alpha)) + \sqrt{\frac{S}{T}}B(\alpha, T)$ by Lemma C.5. It then follows that

$$\sqrt{ST}(\widehat{\alpha} - \alpha) = \psi_1 + \sqrt{\frac{S}{T}}B(\alpha, T) + \psi_2/\sqrt{ST} + \psi_3/ST + R/(ST)^{3/2} \quad (25)$$

for ψ_1, ψ_2, ψ_3 , and R defined above.

ii. Recall

$$\begin{aligned} \widehat{\alpha}^{(\infty)} &= \alpha_T^{-1}(\widehat{\alpha}) \\ &= \alpha_T^{-1}(\alpha_T(\alpha)) + H^{-1}|_{\alpha_T(\alpha)}(\widehat{\alpha} - \alpha_T(\alpha)) \\ &\quad + \sum_{i=1}^p \frac{\partial H^{-1}}{\partial \alpha_i}|_{\alpha_T(\alpha)}(\widehat{\alpha} - \alpha_T(\alpha))(\widehat{\alpha} - \alpha_T(\alpha))_i \\ &\quad + \sum_{i=1}^p \sum_{j=1}^p \frac{\partial^2 H^{-1}}{\partial \alpha_i \partial \alpha_j}|_{\widetilde{\alpha_T(\alpha)}}(\widehat{\alpha} - \alpha_T(\alpha))(\widehat{\alpha} - \alpha_T(\alpha))_i(\widehat{\alpha} - \alpha_T(\alpha))_j \end{aligned} \quad (26)$$

where $\|\widetilde{\alpha_T(\alpha)} - \alpha_T(\alpha)\| \leq \|\widehat{\alpha} - \alpha_T(\alpha)\|$. From Lemma C.5, $\alpha_T(\alpha) = \alpha + \frac{1}{T-p}B(\alpha, T)$, so $H = I + \frac{1}{T-p} \frac{\partial B(\alpha, T)}{\partial \alpha}$, $\frac{\partial H}{\partial \alpha_i} = \frac{1}{T-p} \frac{\partial^2 B(\alpha, T)}{\partial \alpha \partial \alpha_i} = O(1/T)$ by Assumption HO5, and $\frac{\partial^2 H}{\partial \alpha_i \partial \alpha_j} = \frac{1}{T-p} \frac{\partial^3 B(\alpha, T)}{\partial \alpha \partial \alpha_i \partial \alpha_j} = O(1/T)$ by Assumption HO5. Thus, $\frac{\partial H^{-1}}{\partial \alpha_i}|_{\alpha_T(\alpha)} = -H^{-1} \frac{\partial H}{\partial \alpha_i} H^{-1}|_{\alpha_T(\alpha)} = O(1/T)$ and $\frac{\partial^2 H^{-1}}{\partial \alpha_i \partial \alpha_j}|_{\widetilde{\alpha_T(\alpha)}} = H^{-1} \frac{\partial H}{\partial \alpha_i} H^{-1} \frac{\partial H}{\partial \alpha_j} H^{-1} + H^{-1} \frac{\partial H}{\partial \alpha_j} H^{-1} \frac{\partial H}{\partial \alpha_i} H^{-1} - H^{-1} \frac{\partial^2 H}{\partial \alpha_i \partial \alpha_j} H^{-1}|_{\widetilde{\alpha_T(\alpha)}} = O(1/T)$ by Assumption HO5. Also, we have

$$\begin{aligned} H^{-1} &= I - \frac{1}{T-p} \frac{\partial B(\alpha, T)}{\partial \alpha} + \frac{1}{(T-p)^2} \frac{\partial^2 B(\alpha, T)}{\partial \alpha \partial \alpha} \\ &\quad - \widetilde{H}^{-1} \frac{1}{T-p} \frac{\partial B(\alpha, T)}{\partial \alpha} \widetilde{H}^{-1} \frac{1}{T-p} \frac{\partial B(\alpha, T)}{\partial \alpha} \widetilde{H}^{-1} \frac{1}{T-p} \frac{\partial B(\alpha, T)}{\partial \alpha} \widetilde{H}^{-1} \\ &\equiv I + \frac{1}{\sqrt{ST}} H_1 + \frac{1}{ST} H_2 + \frac{1}{(ST)^{3/2}} R_H \end{aligned}$$

where H_1, H_2 , and H_3 are $O(1)$ by Assumption HO5 and $\frac{S}{T} \rightarrow \rho$. Finally plugging the expansion for H^{-1} , the expressions for the derivatives of H and H^{-1} , and (23) into (26) and collecting terms of the same orders yields

$$\begin{aligned} \sqrt{ST}(\widehat{\alpha}^{(\infty)} - \alpha) &= \psi_1 + \psi_2/\sqrt{ST} + H_1|_{\alpha_T(\alpha)}\psi_1/\sqrt{ST} \\ &\quad + \psi_3/ST + H_1|_{\alpha_T(\alpha)}\psi_2/ST + H_2|_{\alpha_T(\alpha)}\psi_1/ST \\ &\quad - \sum_{i=1}^p H^{-1} \sqrt{ST} \frac{\partial^2 B(\alpha, T)}{\partial \alpha \partial \alpha_i} H^{-1}|_{\alpha_T(\alpha)} \psi_1(\psi_1)_i / ST + R^{(\infty)}/(ST)^{3/2} \\ &= \psi_1 + \psi_2^{(\infty)}/\sqrt{ST} + \psi_3^{(\infty)}/ST + R^{(\infty)}/(ST)^{3/2}. \end{aligned} \quad (27)$$

Step 5. Higher Order Bias and Variance. Recall

$$\widehat{\beta}_1(\widetilde{\alpha}) - \beta_1 = \widehat{\beta}_1(\widetilde{\alpha}) - \widehat{\beta}_1(\alpha) + \widehat{\beta}_1(\alpha) - \beta_1. \quad (28)$$

(i) Plugging (17) in for $\widehat{\beta}_1(\widetilde{\alpha}) - \widehat{\beta}_1(\alpha)$ and (25) in for the d_i in (28) yields

$$\begin{aligned} \sqrt{ST}(\widehat{\beta}_1(\widehat{\alpha}) - \beta_1) &= \sqrt{ST}(\widehat{\beta}_1(\alpha) - \beta_1) + \sum_{i=1}^p A^{-1} \Psi_i(\psi_1 + \sqrt{\frac{S}{T}} B(\alpha, T))_i / \sqrt{ST} \\ &\quad + \sum_{i=1}^p A^{-1} \Psi_i(\psi_2)_i / ST \\ &\quad + \frac{1}{2} \sum_{i=1}^p \sum_{j=1}^p (A^{-1} \Psi_{ij} - A^{-1} A_j A^{-1} \Psi_i - A^{-1} A_i A^{-1} \Psi_j) \times \\ &\quad (\psi_1 + \sqrt{\frac{S}{T}} B(\alpha, T))_i (\psi_1 + \sqrt{\frac{S}{T}} B(\alpha, T))_j / ST + R / (ST)^{3/2} \end{aligned}$$

where $R = O_p(1)$. Since the Ψ_i, Ψ_{ij}, \dots are odd functions of V and $\widehat{\alpha}$ is an even function of V , it follows that

$$\text{Bias}(\widehat{\beta}_1(\widehat{\alpha})) = 0.$$

Using the independence obtained in Step 1, the higher order variance is

$$\begin{aligned} \text{Var}(\sqrt{ST}(\widehat{\beta}_1(\widehat{\alpha}) - \beta_1)) &= A^{-1} \\ &\quad + \text{E}[(\sum_{i=1}^p A^{-1} \Psi_i(\psi_1 + \sqrt{\frac{S}{T}} B(\alpha, T))_i / \sqrt{ST})(\sum_{i=1}^p A^{-1} \Psi_i(\psi_1 + \sqrt{\frac{S}{T}} B(\alpha, T))_i / \sqrt{ST})']. \end{aligned}$$

Following Rothenberg (1984), note that since $\widehat{\alpha}$ is an even function of V and S_i is an odd function of V , the ψ 's and Ψ_i are uncorrelated. Hence,

$$\begin{aligned} \text{E}[(\sum_{i=1}^p A^{-1} \Psi_i(\psi_1 + \sqrt{\frac{S}{T}} B(\alpha, T))_i / \sqrt{ST})(\sum_{i=1}^p A^{-1} \Psi_i(\psi_1 + \sqrt{\frac{S}{T}} B(\alpha, T))_i / \sqrt{ST})'] \\ = \frac{1}{ST} \sum_{i=1}^p \sum_{j=1}^p \left(\text{E}[\Psi_i \Psi'_j] \text{E}[(\psi_1)_i (\psi_1)_j] + \frac{S}{T} B(\alpha, T)_i B(\alpha, T)_j \right). \end{aligned}$$

Lemmas C.6 and C.7 and Assumption 4 imply that $\text{E}[\psi_1 \psi'_1] = \frac{ST}{S(T-p)} (\Gamma_p^{-1} + \frac{1}{T} \Gamma_p^{-1} \Theta \Gamma_p^{-1}) = \frac{ST}{S(T-p)} \Gamma_p^{-1} + O(1/T)$. Then defining $\zeta_{ij} = \text{E}[\Psi_i \Psi'_j]$ and

$$\Upsilon = \sum_{i=1}^p \sum_{j=1}^p \zeta_{ij} \left(\frac{ST}{S(T-p)} \Gamma_p^{-1} \right)_{ij} \quad (29)$$

yields

$$\text{Var}(\sqrt{ST}(\widehat{\beta}_1(\widehat{\alpha}) - \beta_1)) = A^{-1} + \Upsilon / ST + \frac{1}{ST} \sum_{i=1}^p \sum_{j=1}^p \zeta_{ij} \frac{S}{T} B(\alpha, T)_i B(\alpha, T)_j + O(1/ST^2). \quad (30)$$

(ii) The derivation with $\hat{\alpha}$ replaced with $\hat{\alpha}^{(\infty)}$ is similar. In particular, the expansion for $\hat{\beta}_1(\hat{\alpha}^{(\infty)} - \beta_1)$ is

$$\begin{aligned} \sqrt{ST}(\hat{\beta}_1(\hat{\alpha}^{(\infty)}) - \beta_1) &= \sqrt{ST}(\hat{\beta}_1(\alpha) - \beta_1) + \sum_{i=1}^p A^{-1}\Psi_i(\psi_1)_i/\sqrt{ST} + \sum_{i=1}^p A^{-1}\Psi_i(\psi_2^{(\infty)})_i/ST \\ &+ \frac{1}{2} \sum_{i=1}^p \sum_{j=1}^p (A^{-1}\Psi_{ij} - A^{-1}A_jA^{-1}\Psi_i - A^{-1}A_iA^{-1}\Psi_j)(\psi_1)_i(\psi_1)_j/ST \\ &+ R/(ST)^{3/2}. \end{aligned}$$

It then follows as in (i) that

$$\text{Bias}(\hat{\beta}_1(\hat{\alpha}^{(\infty)})) = 0$$

and that

$$\text{Var}(\sqrt{ST}(\hat{\beta}_1(\hat{\alpha}^{(\infty)}) - \beta_1)) = A^{-1} + \Upsilon/ST + O(1/ST^2). \quad (31)$$

Finally, the $O(1/ST^2)$ term in (30) and (31) is the same, so

$$\text{Var}(\sqrt{ST}(\hat{\beta}_1(\hat{\alpha}) - \beta_1)) - \text{Var}(\sqrt{ST}(\hat{\beta}_1(\hat{\alpha}^{(\infty)}) - \beta_1)) = \frac{1}{ST} \sum_{i=1}^p \sum_{j=1}^p \zeta_{ij} \frac{S}{T} B(\alpha, T)_i B(\alpha, T)_j$$

is positive semidefinite since

$$\frac{1}{ST} \sum_{i=1}^p \sum_{j=1}^p \zeta_{ij} \frac{S}{T} B(\alpha, T)_i B(\alpha, T)_j = \frac{1}{ST} \frac{S}{T} \mathbb{E} \left[\left(\sum_{i=1}^p B(\alpha, T)_i \Psi_i \right) \left(\sum_{j=1}^p B(\alpha, T)_j \Psi_j' \right) \right].$$

■

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