

## DISTRIBUTED LAG APPROXIMATION TO LINEAR TIME-INVARIANT SYSTEMS<sup>1</sup>

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An infinite distributed lag system is approximated by a truncated one, which can be consistently estimated. We investigate the rate at which the parameter space in the approximate model can be increased as sample size increases, in such a way that these estimates will provide consistent estimates of the underlying system.

**1. Introduction.** Let two time series,  $y(n)$ ,  $z(n)$  observed for  $1 \leq n \leq N$ , and one unobservable time series,  $x(n)$ , be related by

$$(1.1) \quad y(n) = \sum_{j=-\infty}^{\infty} \beta(j)z(n-j) + x(n), \quad n = 0, \pm 1, \dots,$$

where  $Ex(n) = 0$ ,  $Ex(m)z(n) = 0$ , all  $m, n$ . Assuming the means of  $y(n)$  and  $z(n)$  exist and are independent of  $n$ , there is no loss of generality in taking them to be zero also. In practice (1.1) is often replaced by the finite-parameter approximate model

$$(1.2) \quad E(y(n)|z(m), -\infty < m < \infty) = \sum_{j=-p}^q b(j)z(n-j),$$

for some  $p, q$ ,  $0 \leq p, q \leq \infty$ .

The  $\beta(j)$  in (1.1) may be thought of as coefficients in the Laurent expansion of a function  $\tilde{\beta}(s)$ , so

$$\tilde{\beta}(s) = \sum_{j=-\infty}^{\infty} \beta(j)s^j, \quad s \neq 0.$$

From the Weierstrass approximation theorem, it follows that if  $\tilde{\beta}(s)$  is continuous on the circle  $|s| = 1$ , one can choose  $p, q$  and the  $b(j)$  such that, for any  $\epsilon > 0$ ,

$$\max_{|s|=1} |\tilde{b}_{pq}(s) - \tilde{\beta}(s)| < \epsilon, \quad \tilde{b}_{pq}(s) = \sum_{j=-p}^q b(j)s^j.$$

However, in practice,  $p$  and  $q$  are limited by sample size. For estimates based on (1.2) to be reasonably precise,  $p + q$  needs to be substantially smaller than  $N$ . On the other hand, if  $p$  and  $q$  are too small, the discrepancy between (1.1) and (1.2) may be such as to produce serious bias in the estimates. In practice, therefore, the values of  $p$  and  $q$  should depend on  $N$ .

Define by  $b_{Npq}(j)$  a consistent estimate of  $b(j)$ , for given, fixed  $p, q$ . Define also

$$\tilde{b}_{Npq}(s) = \sum_{j=-p}^q b_{Npq}(j)s^j.$$

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The preceding remarks suggest that the construction of a relevant asymptotic theory would include finding functions  $p = p(N)$ ,  $q = q(N)$ , such that, as  $N, p, q \rightarrow \infty$ ,

$$\max_{|s|=1} |\tilde{b}_{Npq}(s) - \tilde{\beta}(s)| \rightarrow 0$$

in probability or with probability one. In this paper we obtain such functions, for estimates  $b_{Npq}(j)$  that are closely related to the least squares estimates. (Our  $q$  sequences apply immediately to the one-sided case  $\beta(j) = 0$ , all  $j < 0$ .) It is likely that very similar results will hold for other estimates. We do not assume  $x(n)$  is white noise so for given  $p$  and  $q$  it may be possible to find more efficient estimates of the  $b(j)$ ; for such estimates when  $x(n)$  is stationary (which we do not assume) see Hannan [4, Chapter VII], Wahba [10]. However, Sims [9] has shown that the asymptotic ( $N \rightarrow \infty$ ) covariance matrices of ordinary and generalized least squares estimates of (1.2) are identical in the limit as  $p, q \rightarrow \infty$ , even when  $x(n)$  is not white noise. Results of this kind supply grounds for using the computationally simplest consistent estimates.

Sims [7], [8] discusses other issues in the approximation of (1.1) by (1.2). Brillinger [3] considers a frequency domain approximation to (1.1). He suggests estimates of  $\tilde{\beta}(e^{i\lambda})$  based on estimates of spectra and cross-spectra, and investigates their asymptotic behavior as  $N \rightarrow \infty$  and a bandwidth parameter  $\rightarrow 0$ . Berk [1], Shibata [6], consider autoregressive estimates of the spectral density, and obtain information on how fast one can increase the autoregressive order as  $N \rightarrow \infty$ .

**2. Main assumptions and definitions.** The initial assumptions are as follows.

CONDITION A.

$$(2.1) \quad z(n) = \sum_{j=-\infty}^{\infty} \theta(j) \zeta(n-j), \quad \sum_{j=-\infty}^{\infty} |\theta(j)| < \infty;$$

$$(2.2) \quad E(\zeta(n) | \zeta(m), m < n; x(m), -\infty < m < \infty) = 0;$$

$$(2.3) \quad E(\zeta(n)^2 | \zeta(m), m < n) = \sigma^2 < \infty;$$

$$(2.4) \quad E|\zeta(n)|^{2\nu} \leq K, \quad \text{some } \nu, \quad 1 < \nu \leq 2.$$

(Throughout,  $K$  represents a finite constant, not necessarily the same one, that may depend on  $\nu$  but is independent of  $p, q, N$  and the  $\zeta(n)$  and  $x(n)$ .)

CONDITION B.  $E|x(n)|^{2\nu} \leq K$ .

CONDITION C.  $\sum_{j=-\infty}^{\infty} |\beta(j)| < \infty$ .

It follows from (2.1)–(2.3) that  $z(n)$  is wide-sense stationary, with

$$c(j) = Ez(n)z(n-j) = \sigma^2 \sum_{k=-\infty}^{\infty} \theta(k)\theta(k-j), \quad -\infty < j < \infty,$$

$|c(j)| \leq K$ . Although  $x(n)$  and  $y(n)$  are not necessarily stationary, Conditions A, B and C imply

$$d(j) = Ey(n)z(n-j) = \sum_{k=-\infty}^{\infty} \beta(k)c(k-j), \quad -\infty < j < \infty,$$

$|d(j)| \leq K$ . Now define the  $(p+q+1) \times (p+q+1)$  matrix  $A_{pq}$ , having  $(j, k)$ th

element  $c(j - k)$ , and the  $(p + q + 1) \times 1$  vector  $a_{pq}$  having  $(j + p + 1)$ th element  $d(j)$ ,  $-p \leq j \leq q$ . Note that if (1.1) were the true model, the  $(p + q + 1) \times 1$  vector  $b_{pq}$ , with  $j$ th element  $b(j - 1 - p)$ , would be identified as  $A_{pq}^{-1}a_{pq}$ .

Now introduce

$$c_{LM}(j) = \frac{1}{M - L} \sum_{n=L+1}^M z(n)z(n - j), \quad d_{LM}(j) = \frac{1}{M - L} \sum_{n=L+1}^M y(n)z(n - j),$$

$$\bar{y}_{LM} = \frac{1}{M - L} \sum_{n=L+1}^M y(n), \quad \bar{z}_{LM} = \frac{1}{M - L} \sum_{n=L+1}^M z(n),$$

for  $0 \leq L < M$ . For  $N > p + q$  define the  $(p + q + 1) \times 1$  vector

$$b_{Npq} = A_{Npq}^{-1}a_{Npq},$$

where  $b_{Npq}$  has  $j$ th element  $b_{Npq}(j - 1 - p)$ ,  $A_{Npq}$  is a  $(p + q + 1) \times (p + q + 1)$  Toeplitz matrix with  $(j, k)$ th element ( $j \geq k$ )  $c_{j-k, N}(j - k) - \bar{z}_{0N}^2$ , and  $a_{Npq}$  has  $(j + p + 1)$ th element  $d_{0, N+j}(j) - \bar{y}_{0N}\bar{z}_{0N}$  for  $-p \leq j < 0$ , and  $d_{jN}(j) - \bar{y}_{0N}\bar{z}_{0N}$ , for  $0 \leq j \leq q$ .

**3. Six lemmas.** Lemma 1 will not be proved as it is a special case of Lemma 1 of Robinson [5].

LEMMA 1. For  $L < M$ , let

$$(3.1) \quad S_{LM} = \frac{1}{M - L} \sum_{L+1}^M u(n), \quad u(n) = \sum_{l, m=-\infty}^{\infty} v_{lm}(n),$$

$$E(v_{lm}(n)|v_{lm}(n - 1), v_{lm}(n - 2), \dots) = 0, \quad \text{a.s.},$$

for all  $l, m, n$ . For some  $\nu$ ,  $1 < \nu \leq 2$ , let there exist  $\eta_1(l), \eta_2(l)$ ,  $-\infty < l < \infty$ , such that

$$(3.2) \quad E|v_{lm}(n)|^p \leq |\eta_1(l)\eta_2(m)|^p, \quad \sum_{l=-\infty}^{\infty} |\eta_j(l)| < \infty, \quad j = 1, 2,$$

for all  $l, m, n$ . Then

$$E|S_{LM}|^p \leq K(M - L)^{1-\nu}.$$

In the next two lemmas, we derive bounds on the central moments of  $c_{LM}(j)$  and  $d_{LM}(j)$  that are independent of  $j$ .

LEMMA 2. Under Condition A,

$$E|c_{LM}(j) - c(j)|^p \leq K(M - L)^{1-\nu}.$$

PROOF. Note that  $c_{LM}(j) - c(j)$  has the form of  $S_{LM}$  in Lemma 1 if  $v_{lm}(n) = \theta(l)\theta(l - j)(\zeta(n - l)^2 - \sigma^2)$ ,  $m = l - j$ ;  $= \theta(l)\theta(m)\zeta(n - l)\zeta(n - j - m)$ ,  $m \neq l - j$ . To see this, observe that (3.1) is true because (2.2), (2.3) imply

$$E(\zeta(n)^2 - \sigma^2|\zeta(n - l)^2 - \sigma^2, l > 0) = 0,$$

$$E(\zeta(n)\zeta(m)|\zeta(n - l)\zeta(m - l), l > 0)$$

$$= E(E(\zeta(n)\zeta(m)|\zeta(n - l), l > 0)|\zeta(n - l)\zeta(m - l), l > 0)$$

$$= E(\zeta(m)E(\zeta(n)|\zeta(n - l), l > 0)|\zeta(n - l)\zeta(m - l), l > 0)$$

$$= 0, \quad m < n.$$

Also, (3.2) is true because, by (2.4),

$$E|v_{lm}(n)|^p \leq |\theta(l)\theta(l-j)|^p 2^p (E|\zeta(n-l)|^{2p} + \sigma^{2p}) \leq K|\theta(l)\theta(l-j)|^p, \quad m = l - j,$$

$$E|v_{lm}(n)|^p \leq |\theta(l)\theta(m)|^p (E|\zeta(n-l)|^{2p} E|\zeta(n-m)|^{2p})^{\frac{1}{2}} \leq K|\theta(l)\theta(m)|^p, \quad m \neq l - j.$$

Then the result follows from Lemma 1.  $\square$

LEMMA 3. Under Conditions A, B and C,

$$E|d_{LM}(j) - d(j)|^p \leq K(M - L)^{1-p}.$$

PROOF. Write

$$\begin{aligned} d_{LM}(j) - d(j) &= e_{LM}(j) + f_{LM}(j), \\ e_{LM}(j) &= \sum_{k=-\infty}^{\infty} \beta(k)(c_{L-k, M-k}(j-k) - c(j-k)) \\ f_{LM}(j) &= \frac{1}{M-L} \sum_{n=L+1}^M x(n)z(n-j). \end{aligned}$$

Now by the  $c_r$ -inequality and Jensen's inequality,

$$(3.3) \quad E|d_{LM}(j) - d(j)|^p \leq 2E|e_{LM}(j)|^p + 2E|f_{LM}(j)|^p.$$

By Hölder's inequality and Lemma 2,

$$\begin{aligned} E|e_{LM}(j)|^p &\leq (\sum |\beta(k)|)^{p-1} \sum |\beta(k)| E|c_{L-k, M-k}(j-k) - c(j-k)|^p \\ &\leq K(\sum |\beta(k)|)^p (M - L)^{1-p} = K(M - L)^{1-p}. \end{aligned}$$

To handle  $f_{LM}(j)$  take

$$v_{lm}(n) = \theta(l)\zeta(n-l)x(n), \quad m = 0; = 0, m \neq 0$$

in Lemma 1 and note that by (2.2), (2.4) and Condition B,

$$\begin{aligned} &E(\zeta(m)x(n)|\zeta(m-l)x(n-l), l > 0) \\ &= E(x(n)E(\zeta(m)|\zeta(m-l), l > 0; x(l), -\infty < l < \infty)|\zeta(m-l)x(n-l), l > 0) = 0, \\ &E|\zeta(m)x(n)|^p \leq (E|\zeta(m)|^{2p} E|x(n)|^{2p})^{\frac{1}{2}} \leq K. \end{aligned}$$

It follows that  $E|f_{LM}(j)|^p < K(M - L)^{1-p}$ . Finally, apply (3.3).  $\square$

From Lemma 2 of [5] and Conditions A-C it may readily be shown also that

$$\begin{aligned} (3.4) \quad &E|\bar{z}_{LM}|^{2p} \leq K(M - L)^{-p} \\ &E|\bar{y}_{LM}|^{2p} \leq E|\sum \beta(k)\bar{z}_{L-k, M-k} + \bar{x}_{LM}|^{2p} \\ &\leq 2^p \{ (\sum |\beta(k)|)^{2p-1} \sum |\beta(k)| E|\bar{z}_{L-k, M-k}|^{2p} + E|\bar{x}_{LM}|^{2p} \} \\ (3.5) \quad &\leq K(M - L)^{-p}, \end{aligned}$$

where  $\bar{x}_{LM} = (M - L)^{-1}(x(L + 1) + \dots + x(M))$ .

Let  $\|X\|$  be the square root of the greatest eigenvalue of  $XX'$ , where  $X'$  is the transposed of  $X$ . Define  $\Delta_{Npq} = A_{Npq} - A_{pq}$ .

LEMMA 4. Under Condition A,

$$E\|\Delta_{Npq}\|^p \leq K(p + q + 1)^p(N - p - q)^{1-p}, \quad N > p + q.$$

PROOF. Because  $\Delta_{Npq}$  is symmetric,  $\|\Delta_{Npq}\|$  is its greatest eigenvalue. Then from a theorem of Perron

$$(3.6) \quad \|\Delta_{Npq}\| \leq \max_{0 \leq j < p+q} \sum_{k=0}^{p+q} |c_{|j-k|, N}(|j-k|) - \bar{z}_{0N}^2 - c(j-k)| \\ \leq 2 \sum_{j=0}^{p+q} (|c_{jN}(j) - c(j)| + \bar{z}_{0N}^2),$$

the second inequality resulting because  $\Delta_{Npq}$  is Toeplitz. Thus from Lemma 2, (3.4) and Minkowski's inequality

$$E\|\Delta_{Npq}\|^p \leq 4(p + q + 1)^{p-1} \sum_{j=0}^{p+q} \{ E|c_{jN}(j) - c(j)|^p + E|\bar{z}_{0N}|^{2p} \} \\ \leq K(p + q + 1)^{p-1} \sum_{j=0}^{p+q} \{ (N - j)^{1-p} + N^{-p} \} \\ \leq K(p + q + 1)^p(N - p - q)^{1-p}. \quad \square$$

Define  $d_{Npq} = a_{Npq} - a_{pq}$ .

LEMMA 5. Under Conditions A, B and C,

$$E\|\Delta_{Npq}\|^p \leq K(p + q + 1)^p(N - p - q)^{1-p}, \quad N > p + q.$$

PROOF. In this case

$$E\|\delta_{Npq}\|^p \leq E \{ \sum_{j=-p}^{-1} |d_{0, N+j}(j) - \bar{y}_{0N}\bar{z}_{0N} - d(j)| \\ + \sum_{j=0}^q |d_{jN}(j) - \bar{y}_{0N}\bar{z}_{0N} - d(j)| \}^p \\ \leq 4p^{p-1} \sum_{j=-p}^{-1} E|d_{0, N+j}(j) - d(j)|^p + 4(q + 1)^{p-1} \sum_0^q E|d_{jN}(j) - d(j)|^p \\ + 2(p + q + 1)^p \{ E|\bar{y}_{0N}|^{2p} E|\bar{z}_{0N}|^{2p} \}^{\frac{1}{2}} \\ \leq K \{ p^p(N - p)^{1-p} + (q + 1)^p(N - q)^{1-p} + (p + q + 1)^p N^{-p} \} \\ \leq K(p + q + 1)^p(N - p - q)^{1-p},$$

using Lemma 3 and (3.4), (3.5).  $\square$

For the final lemma we introduce

CONDITION D.  $\sum_{j=-\infty}^{\infty} \theta(j)s^j$  is bounded away from zero on  $|s| = 1$ .

This is equivalent to the power spectrum of  $z(n)$  (which exists under A) being bounded away from zero.

LEMMA 6. Under Condition D,

$$\|A_{pq}^{-1}\| \leq K.$$

PROOF. If  $\|A_{pq}^{-1}\|$  exists it is the reciprocal of the smallest eigenvalue,  $\rho$ , of  $A_{pq}$ .

Thus for some real numbers  $w_j$ ,  $0 \leq j \leq p + q$ ,  $w_0^2 + \dots + w_{p+q}^2 = 1$ ,

$$\begin{aligned} \rho &= \sum_{j,k=0}^{p+q} \sum w_j c(j-k) w_k = \sigma^2 \sum_{j,k=0}^{p+q} \sum \sum_{l=-\infty}^{\infty} w_j \theta(l) \theta(l+j-k) w_k \\ &= \sigma^2 \int_{|s|=1} |\sum_{j=0}^{p+q} w_j s^j|^2 |\sum_{j=-\infty}^{\infty} \theta(j) s^j|^2 ds > K \int_{|s|=1} |\sum_{j=0}^{p+q} w_j s^j|^2 ds, \end{aligned}$$

by Condition D. The last expression is  $2\pi K(w_0^2 + \dots + w_{p+q}^2) = 2\pi K > 0$ , so  $\rho^{-1} \leq K$ .  $\square$

**4. Weak convergence.** We introduce nonnegative integers  $u, v$  and real values  $\xi, \chi$ ,  $0 \leq \xi, \chi < 1$ , which describe the smoothness of  $\tilde{\beta}(s)$  as follows:

CONDITION E.  $\sum_{j=-\infty}^0 \beta(j) s^j$  is  $u$ -times differentiable on  $|s| = 1$ , its  $u$ th derivative satisfying a Lipschitz condition of order  $\xi$ ;  $\sum_{j=1}^{\infty} \beta(j) s^j$  is  $v$ -times differentiable on  $|s| = 1$ , its  $v$ th derivative satisfying a Lipschitz condition of order  $\chi$ .

This condition implies C when  $\xi > \frac{1}{2}$  and  $\chi > \frac{1}{2}$  (Zygmund [11, page 64]).

Let  $\beta_{pq}$  be the  $(p + q + 1) \times 1$  vector with  $j$ th element  $\beta(j - 1 - p)$ .

**THEOREM 1.** *Let Conditions A, B, D and E hold, with  $u + \xi > \frac{1}{2}$ ,  $v + \chi > \frac{1}{2}$ .*

*Then*

$$(4.1) \quad p \lim_{N,p,q \rightarrow \infty} \|b_{Npq} - \beta_{pq}\| = 0,$$

*if*

$$(4.2) \quad q \leq Kp^{2(u+\xi)-\epsilon}, \quad p \leq Kq^{2(v+\chi)-\epsilon}, \quad \text{any } \epsilon > 0,$$

$$(4.3) \quad (p + q) / N^{(v-1)/v} \rightarrow 0.$$

**PROOF.** Write

$$\begin{aligned} b_{Npq} - \beta_{pq} &= A_{Npq}^{-1} (a_{Npq} - A_{Npq} \beta_{pq}) \\ &= A_{Npq}^{-1} (\delta_{Npq} - \Delta_{Npq} \beta_{pq} + a_{pq} - A_{pq} \beta_{pq}) \end{aligned}$$

so that

$$(4.4) \quad \|b_{Npq} - \beta_{pq}\| \leq \|A_{Npq}^{-1}\| (\|\delta_{Npq}\| + \|\Delta_{Npq}\| \|\beta_{pq}\| + \|a_{pq} - A_{pq} \beta_{pq}\|).$$

Now

$$\|A_{Npq}^{-1}\| \leq \|A_{Npq}^{-1} - A_{pq}^{-1}\| + \|A_{pq}^{-1}\| \leq (\|A_{Npq}^{-1}\| \|\Delta_{Npq}\| + 1) \|A_{pq}^{-1}\|,$$

so

$$\|A_{Npq}^{-1}\| (1 - \|A_{pq}^{-1}\| \|\Delta_{Npq}^N\|) \leq \|A_{pq}^{-1}\|.$$

By Markov's inequality it follows from Lemma 4 that  $\|\Delta_{Npq}\| \rightarrow 0$ , in probability, if

$$(p + q + 1)^v (N - p - q)^{1-v} \rightarrow 0, \quad \text{i.e., } (p + q)^v N^{1-v} \left(1 - \frac{p + q}{N}\right)^{1-v} \rightarrow 0.$$

Thus from Lemma 6

$$p \lim_{N,p,q \rightarrow \infty} \|A_{Npq}^{-1}\| \leq \lim_{p,q \rightarrow \infty} \|A_{pq}^{-1}\| < \infty,$$

for sequences (4.3). We are left with the factor in parentheses in (4.4). By Lemma 5

and Markov's inequality,  $\|\delta_{Npq}\| \rightarrow 0$  i.p. for sequences (4.3). The second term  $\rightarrow 0$  i.p. since  $\|\Delta_{Npq}\| \rightarrow 0$  i.p. and

$$\|\beta_{pq}\| = \left(\sum_{j=-p}^q \beta(j)^2\right)^{\frac{1}{2}} \leq \sum_{j=-p}^q |\beta(j)| \leq \sum_{j=-\infty}^{\infty} |\beta(j)| < \infty,$$

by *E* and  $u + \xi, v + \chi > \frac{1}{2}$ . Finally, for  $p$  and  $q$  that are not too small,

$$\begin{aligned} \|a_{pq} - A_{pq}\beta_{pq}\| &= \left\{ \sum_{j=-p}^q (d(j) - \sum_{k=-p}^q \beta(k)c(j+k))^2 \right\}^{\frac{1}{2}} \\ &= \left\{ \sum_{j=-p}^q \left( \sum_{k=-\infty}^{-p-1} \beta(k)c(j+k) + \sum_{k=q+1}^{\infty} \beta(k)c(j+k) \right)^2 \right\}^{\frac{1}{2}} \\ &= \frac{\sigma^2}{2\pi} \left\{ \sum_{-p}^q \int_{|s|=1} \left( \sum_{-\infty}^{-p-1} \beta(k)s^k + \sum_{q+1}^{\infty} \beta(k)s^k \right) s^j \left| \sum_{-\infty}^{\infty} \theta(l)s^l \right|^2 ds \right\}^{\frac{1}{2}} \\ &< c(0)(p+q+1)^{\frac{1}{2}} \left\{ \max_{|s|=1} \left| \sum_{-\infty}^{-p-1} \beta(k)s^k \right| + \max_{|s|=1} \left| \sum_{q+1}^{\infty} \beta(k)s^k \right| \right\} \\ &< K(p^{\frac{1}{2}} + q^{\frac{1}{2}}) \left( \frac{\ln p}{p^{u+\xi}} + \frac{\ln q}{q^{v+\chi}} \right) \end{aligned}$$

(Zygmund [11, page 120]). From  $u + \xi, v + \chi > \frac{1}{2}$  and (4.2), this  $\rightarrow 0$  as  $p, q \rightarrow \infty$ .  $\square$

When  $\beta(j) = 0$ , all  $j < J$ , some  $J > -\infty$ , the conditions for (4.1) are that *A, B, D*, and *E* hold, with  $v + \chi > \frac{1}{2}$ , and  $qN^{(v-1)/v} \rightarrow 0$  as  $N \rightarrow \infty$ .

**THEOREM 2.** *Let Conditions A, B, D and E hold, with  $u + \xi > 1, v + \chi > 1$ . Then*

$$p \lim_{N,p,q \rightarrow \infty} \max_{|s|=1} |\tilde{b}_{Npq}(s) - \tilde{\beta}(s)| = 0,$$

if

$$(4.5) \quad q \leq Kp^{u+\xi-\epsilon}, p \leq Kq^{v+\chi-\epsilon}, \quad \text{any } \epsilon > 0,$$

$$(4.6) \quad (p+q)/N^{(2(v-1)/3v)} \rightarrow 0.$$

**PROOF.** By the triangle inequality

$$\max_{|s|=1} |\tilde{b}_{Npq}(s) - \tilde{\beta}(s)| \leq \sum_{j=-p}^q |b_{Npq}(j) - \beta(j)| + \sum_{j=-\infty}^{-p-1} |\beta(j)| + \sum_{j=q+1}^{\infty} |\beta(j)|.$$

The last two terms on the right  $\rightarrow 0$  as  $p, q \rightarrow \infty$ . The first term is bounded by  $(p+q+1)^{\frac{1}{2}} \|b_{Npq} - \beta_{pq}\|$ , from the Schwarz inequality. Thus the theorem is proved if

$$\begin{aligned} p \lim_{N,p,q \rightarrow \infty} (p+q)^{\frac{1}{2}} \|\Delta_{Npq}\| &= p \lim_{N,p,q \rightarrow \infty} (p+q)^{\frac{1}{2}} \|\delta_{Npq}\| \\ &= \lim_{p,q \rightarrow \infty} (p+q)^{\frac{1}{2}} \|a_{pq} - A_{pq}\beta_{pq}\| = 0. \end{aligned}$$

The first two limits are zero under (4.5), from Lemmas 4 and 5. Finally, as in the proof of Theorem 1,

$$(p+q)^{\frac{1}{2}} \|a_{pq} - A_{pq}\beta_{pq}\| \leq K(p+q) \left( \frac{\ln p}{p^{u+\xi}} + \frac{\ln q}{q^{v+\chi}} \right) \rightarrow 0,$$

from Zygmund [11, page 120],  $u + \xi, v + \chi > 1$ , and (4.5).

Conditions (4.2), (4.3), (4.5) and (4.6) limit the ultimate rate of increase of  $p$  and  $q$  relative to one another, and to  $N$ .

**5. Strong convergence.** The mode of convergence can be strengthened if the growth of  $p$  and  $q$  is further restricted, to a small degree.

**THEOREM 3.** *Let Conditions A, B, D and E hold, with  $u + \xi > \frac{1}{2}$ ,  $v + \chi > \frac{1}{2}$ . Then*

$$\lim_{N,p,q \rightarrow \infty} \|b_{Npq} - \beta_{pq}\| = 0, \quad \text{a.s.,}$$

under (4.2), and the conditions

$$(5.1) \quad p + q \leq C2^l, \quad \text{some } C < 1,$$

where  $l$  is an integer such that  $2^l \leq N$ , and

$$(5.2) \quad p + q \leq KN^{(\nu-1)/\nu} / \{(\ln N)^{(\nu+1)/\nu} \ln \ln N\}.$$

**PROOF.** The proof follows that of Theorem 1 as soon as we prove

$$\lim_{N,p,q \rightarrow \infty} \|\Delta_{Npq}\| = \lim_{N,p,q \rightarrow \infty} \|\delta_{Npq}\| = 0, \quad \text{a.s.}$$

To establish the first limit, consider (3.6). The term  $2(p + q + 1)\bar{z}_{0N}^2$  in this has  $\nu$ th moment

$$K \left( \frac{p + q + 1}{N} \right)^\nu \leq \frac{K}{N(\ln N)^{\nu+1}(\ln \ln N)^\nu},$$

under (5.2), and thus  $\rightarrow 0$  a.s. by Markov's inequality and the Borel-Cantelli lemma. The other term in (3.6) is twice

$$\sum_{j=0}^{p+q} |c_{jN}(j) - c(j)| \leq \sum_{j=0}^{p+q} \left\{ \left( \frac{L-j}{N-j} \right) |c_{jL}(j) - c(j)| + \left( \frac{N-L}{N-j} \right) |c_{LN}(j) - c(j)| \right\} \\ \leq g_L + h_L,$$

$$g_L = \frac{L}{L-p-q} \sum_{j=0}^{p+q} |c_{jL}(j) - c(j)|,$$

$$h_L = \frac{L}{L-p-q} \max_{L < N < 2L} \sum_{j=0}^{p+q} |c_{LN}(j) - c(j)|,$$

for  $p + q < L < N < 2L$ . Consider  $g_L$ . For  $N$  not too small,

$$Eg_L^\nu \leq (p + q + 1)^{\nu-1} L(L-p-q)^{-\nu} \sum_{j=0}^{p+q} E|c_{jL}(j) - c(j)|^\nu \\ \leq K(p + q + 1)^\nu L^\nu (L-p-q)^{1-2\nu} \\ \leq K \{L / (L-p-q)\}^{2\nu-1} \{(\ln L)^{\nu+1} (\ln \ln L)^\nu\}^{-1},$$

from Lemma 2 and (5.2). Now if we choose  $L = 2^l$ ,  $l$  integer, and use (5.1), the last expression is bounded by  $Kl^{-\nu}$ . Thus by Markov's inequality and the Borel-



Cantelli lemma,  $g_L \rightarrow 0$  a.s. For  $h_L$ ,

$$\begin{aligned} Eh_L^p &\leq (p+q+1)^{p-1} L^p (L-p-q)^{-p} \sum_{j=0}^{p+q} E \{ \max_{L < N < 2L} |c_{LN}(j) - c(j)| \}^p \\ &\leq K(p+q+1)^p L^p (L-p-q)^{-p} (\log_2 4L)^p L^{1-p} \\ &\leq K(\log_2 4L)^p \{(\ln L)^{p+1} (\ln \ln L)^p\}^{-1} \leq Kl^{-1} (\ln l)^{-p}, \end{aligned}$$

using Billingsley [2, page 102], Lemma 2, (5.1) and (5.2). Thus  $h_L \rightarrow 0$  a.s. by Markov's inequality and the Borel-Cantelli lemma, completing the proof that  $\|\Delta_{Npq}\| \rightarrow 0$  a.s. The proof that  $\|\delta_{Npq}\| \rightarrow 0$  a.s. is very similar so we omit it.  $\square$

**THEOREM 4.** *Let Conditions A, B, D and E hold with  $u + \xi > 1$ ,  $v + \chi > 1$ . Then*

$$\lim_{N, p, q \rightarrow \infty} \max_{|s|=1} |\tilde{b}_{Npq}(s) - \tilde{\beta}(s)| = 0, \quad \text{a.s.,}$$

under (4.5), (5.1) and

$$p + q \leq KN^{(2(\nu-1)/3\nu)} / \{(\ln N)^{(2(\nu+1)/3\nu)} (\ln \ln N)^{\frac{2}{3}}\}.$$

The proof is omitted because it uses Theorem 3 in precisely the way the proof of Theorem 2 used Theorem 1.

#### REFERENCES

- [1] BERK, K. N. (1974). Consistent autoregressive spectral estimates. *Ann. Statist.* **2** 489–502.
- [2] BILLINGSLEY, P. (1968). *Convergence of Probability Measures*. Wiley, New York.
- [3] BRILLINGER, D. R. (1975). *Time Series. Data Analysis and Theory*. Holt, Rinehart and Winston, New York.
- [4] HANNAN, E. J. (1970). *Multiple Time Series*. Wiley, New York.
- [5] ROBINSON, P. M. (1978). On consistency in time series analysis. *Ann. Statist.* **6** 215–223.
- [6] SHIBATA, R. (1978). Convergence of least square estimates of autoregressive parameters. *Austral. J. Statist.* **20**.
- [7] SIMS, C. A. (1971). Distributed lag estimation when the parameter space is explicitly infinite-dimensional. *Ann. Math. Statist.* **42** 1622–1632.
- [8] SIMS, C. A. (1972). The role of approximate prior restrictions in distributed lag estimation. *J. Amer. Statist. Assoc.* **67** 164–175.
- [9] SIMS, C. A. (1974). Distributed lags. In *Frontiers of Quantitative Economics, Vol. II* (M. D. Intriligator and D. A. Kendrick, eds.). North-Holland, Amsterdam.
- [10] WAHBA, G. (1969). Estimation of the coefficients in a multidimensional distributed lag model. *Econometrica* **37** 398–407.
- [11] ZYGMUND, A. (1959). *Trigonometric Series Vol. 1*. Cambridge Univ. Press, Cambridge.

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