

## Almost Sure Convergence of Weighted Quadratic Forms for Martingale Difference Sequences

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Received August 10, 2001

Revised January 19, 2002

**Abstract.** The paper is concerned with the study of almost sure convergence of weighted quadratic forms for martingale difference sequences of random variables.

### 1. Introduction

Let  $(\Omega, \mathfrak{F}, P)$  be a probability space and  $\{\mathfrak{F}_n, n \geq 1\}$  be an increasing sequence of sub-sigma fields of  $\mathfrak{F}$  and  $\{X_n, n \geq 1\}$  be a sequence of independent real random variables adapted to  $\{\mathfrak{F}_n, n \geq 1\}$  such that  $E[X_k] = 0$  and  $E[X_k^2] = 1$  for  $k = 1, 2, \dots$ . Let  $(a_{jk}), j, k = 1, 2, \dots$  be a matrix of real numbers (not necessarily symmetric). Varberg (1968) has studied the almost sure convergence of  $S_n = \sum_{j,k=1}^n a_{jk} X_j X_k$  as stated below.

**Theorem A.** *If  $\sum_{j,k=1}^{\infty} a_{jk}^2 < \infty$  and  $\sum_{k=1}^{\infty} |a_{kk}| < \infty$ , then  $S_n = \sum_{j,k=1}^n a_{jk} X_j X_k$  converges almost surely.*

**Theorem B.** *If there are positive constants  $\delta$  and  $\varepsilon$  such that  $P[|X_k| > \delta] \geq \varepsilon, k = 1, 2, \dots$ , if  $\sum_{k=1}^{\infty} |a_{kk}| < \infty$ , then*

(a)  $\sum_{k=1}^n \left[ \sum_{j=1}^{k-1} (a_{kj} + a_{jk}) X_j \right]^2,$

(b)  $\sum_{k=1}^n \left[ \sum_{j=1}^{k-1} (a_{kj} + a_{jk}) X_j \right]^2 X_k^2$  and

(c)  $\sum_{k=1}^n \left[ \sum_{j=1}^{k-1} (a_{kj} + a_{jk}) X_j \right] X_k + \sum_{k=1}^n a_{kk} X_k^2$  converge on equivalent sets, i.e. on sets which differs at most by null sets.

**Definition 1.** An adapted sequence of random variables  $\{X_n, \mathfrak{F}_n, n \geq 1\}$  is said to be a martingale difference sequence if  $E[X_n | \mathfrak{F}_{n-1}] = 0$  a.s. for each  $n \geq 2$ .

In this paper we have proved results of the type mentioned in Theorems B and A for martingale difference sequences.

## 2. Main Results

**Theorem 2.1.** Let

- (i)  $\{X_k, k \geq 1\}$  be a martingale difference sequence with  $E[X_k^4 | \mathfrak{F}_{k-1}] = c < \infty$ ,  $E[X_k^3 | \mathfrak{F}_{k-1}] = 0$ , and  $E[X_k^2 | \mathfrak{F}_{k-1}] = 1$  for all  $k \geq 1$ ;
- (ii)  $(a_{jk}), j, k = 1, 2, \dots$  be a real symmetric matrix with  $\sum_{j,k=1}^{\infty} a_{jk}^2 < \infty$  and  $\sum_{k=1}^{\infty} |a_{kk}| < \infty$ . Then  $S_n = \sum_{j,k=1}^n a_{jk} X_j X_k$  converges in quadratic mean and almost surely.

*Remark 2.1.* If  $(a_{jk})$  is not symmetric,  $a_{jk} = a_j b_k$  and if  $\sum_{j,k=1}^{\infty} a_{jk}^2 = M < \infty$ , then

$$S_n = \sum_{j,k=1}^n a_{jk} X_j X_k = \left( \sum_{j=1}^n a_j X_j \right) \left( \sum_{k=1}^n b_k X_k \right)$$

and

$$T_n = S_n - \sum_{k=1}^n a_{kk} = S_n - \sum_{k=1}^n a_k b_k$$

converge almost surely and we do not need the assumption that  $E(X_n^3 | \mathfrak{F}_{n-1}) = 0$  and  $E(X_n^4 | \mathfrak{F}_{n-1}) < \infty$  for a martingale difference sequence  $\{X_n, n \geq 1\}$ , where  $\sum_{j,k=1}^{\infty} a_{jk}^2 < \infty$ .

**Theorem 2.2.** Let

- (i)  $(a_{jk}), j, k=1, 2, \dots$  be a matrix of real numbers with  $\sum_{j,k=1}^{\infty} a_{jk}^2 < \infty$  and  $\sum_{k=1}^{\infty} |a_{kk}| < \infty$ ;
- (ii)  $\{X_k, k \geq 1\}$  be a martingale difference sequence with  $E[X_k^2 | \mathfrak{F}_{k-1}] = 1$  for all  $k \geq 1$ . Then  $S_n = \sum_{j,k=1}^n a_{jk} X_j X_k$  converges almost surely.

**Theorem 2.3.** Let

- (i)  $(a_{jk}), j, k = 1, 2, \dots$  be a matrix of real numbers with  $\sum_{j,k=1}^{\infty} a_{jk}^2 < \infty$  and  $\sum_{k=1}^{\infty} |a_{kk}| < \infty$ ;
- (ii)  $\{X_k, k \geq 1\}$  be a martingale difference sequence with  $E[X_k^2 | \mathfrak{F}_{k-1}] = 1$  for all  $k \geq 1$  and
- (iii)  $V_k = \sum_{j=1}^{k-1} (a_{kj} + a_{jk}) X_j$ . Then  $\sum_{k=1}^n V_k X_k + \sum_{k=1}^n a_{kk} X_k^2$ ,  $\sum_{k=1}^n V_k^2$ , and  $\sum_{k=1}^n V_k^2 X_k^2$  converge almost surely.

*Proof of Theorem 2.1.* Let  $T_0 = 0$ ,  $T_n = S_n - E[S_n]$  and  $Y_n = T_n - T_{n-1}$  for  $n = 1, 2, \dots$

Now

$$\begin{aligned}
 T_n &= S_n - E[S_n] = \sum_{j,k=1}^n a_{jk}X_jX_k - E\left[\sum_{j,k=1}^n a_{jk}X_jX_k\right] \\
 &= \sum_{j,k=1}^n a_{jk}X_jX_k - E\left[\sum_{k=1}^n a_{kk}X_k^2\right] - E\left[\sum_{j \neq k=1}^n a_{jk}X_jX_k\right] \\
 &= \sum_{j,k=1}^n a_{jk}X_jX_k - \sum_{k=1}^n a_{kk}E[E(X_k^2|\mathfrak{S}_{k-1})] - \sum_{j \neq k=1}^n a_{jk}E[E(X_jX_k|\mathfrak{S}_{\min(j,k)})] \\
 &= \sum_{j,k=1}^n a_{jk}X_jX_k - \sum_{k=1}^n a_{kk} - \sum_{j \neq k=1}^n a_{jk}E[X_{\min(j,k)}E(X_{\max(j,k)}|\mathfrak{S}_{\min(j,k)})] \\
 &= \sum_{j,k=1}^n a_{jk}X_jX_k - \sum_{k=1}^n a_{kk}.
 \end{aligned}$$

Hence

$$S_n = T_n + \sum_{k=1}^n a_{kk}.$$

Again

$$\sum_{k=1}^n Y_k = \sum_{k=1}^n (T_k - T_{k-1}) = T_n.$$

For  $n \neq m$ ,

$$\begin{aligned}
 E[Y_n Y_m] &= E[(T_n - T_{n-1})(T_m - T_{m-1})] \\
 &= E\left\{\left[\sum_{j,k=1}^n a_{jk}X_jX_k - \sum_{k=1}^n a_{kk} - \sum_{j,k=1}^{n-1} a_{jk}X_jX_k + \sum_{k=1}^{n-1} a_{kk}\right]\right. \\
 &\quad \left. \left[\sum_{k,j,k=1}^m a_{jk}X_jX_k - \sum_{k=1}^m a_{kk} - \sum_{j,k=1}^{n-1} a_{jk}X_jX_k + \sum_{k=1}^{m-1} a_{kk}\right]\right\} \\
 &= E\left\{\left[2X_n \sum_{k=1}^{n-1} a_{nk}X_k + a_{nn}(X_n^2 - 1)\right]\left[2X_m \sum_{k=1}^{m-1} a_{mk}X_k + a_{mm}(X_m^2 - 1)\right]\right\} \\
 &= 4E\left[X_n X_m \left(\sum_{k=1}^{n-1} a_{nk}X_k\right) \left(\sum_{k=1}^{m-1} a_{mk}X_k\right)\right] + 2a_{nn}E\left[(X_m^2 - 1)X_m \left(\sum_{k=1}^{m-1} a_{mk}X_k\right)\right] \\
 &\quad + 2a_{mm}E\left[(X_m^2 - 1)X_n \left(\sum_{k=1}^{n-1} a_{nk}X_k\right)\right] + a_{nn}a_{mm}E\left[(X_n^2 - 1)(X_m^2 - 1)\right] \\
 &= 4E\left[E\left\{X_n X_m \left(\sum_{k=1}^{n-1} a_{nk}X_k\right) \left(\sum_{k=1}^{m-1} a_{mk}X_k\right) \mid \mathfrak{S}_{n-1}\right\}\right] \\
 &\quad + 2a_{nn}E\left[E\left\{(X_n^2 - 1)X_m \left(\sum_{k=1}^{m-1} a_{mk}X_k\right) \mid \mathfrak{S}_{n-1}\right\}\right] \\
 &\quad + 2a_{mm}E\left[E\left\{(X_m^2 - 1)X_n \left(\sum_{k=1}^{n-1} a_{nk}X_k\right) \mid \mathfrak{S}_{n-1}\right\}\right]
 \end{aligned}$$

$$\begin{aligned}
& + a_{nn}a_{mm}E[E\{(X_n^2 - 1)(X_m^2 - 1) | \mathfrak{S}_{n-1}\}] \\
= & 4E\left[X_m\left(\sum_{k=1}^{n-1} a_{nk}X_k\right)\left(\sum_{k=1}^{m-1} a_{mk}X_k\right)E\{X_n | \mathfrak{S}_{n-1}\}\right] \\
& + 2a_{nn}E\left[X_m\left(\sum_{k=1}^{m-1} a_{mk}X_k\right)E\{(X_n^2 - 1) | \mathfrak{S}_{n-1}\}\right] \\
& + 2a_{mm}E\left[(X_m^2 - 1)\left(\sum_{k=1}^{n-1} a_{nk}X_k\right)E\{X_n | \mathfrak{S}_{n-1}\}\right] \\
& + a_{nn}a_{mm}E[(X_m^2 - 1)E\{(X_n^2 - 1) | \mathfrak{S}_{n-1}\}] = 0.
\end{aligned}$$

(since  $X_n$  is a martingale difference sequence and  $E[X_n^2 | \mathfrak{S}_{n-1}] = 1$ ) Hence  $Y_n$  is orthogonal. Again

$$Y_n = T_n - T_{n-1} = 2X_n \sum_{k=1}^{n-1} a_{nk}X_k + a_{nn}(X_n^2 - 1) = K_n + L_n \quad (\text{say})$$

Now

$$\begin{aligned}
E[K_n^2] & = E\left[2X_n \sum_{k=1}^{n-1} a_{nk}X_k\right]^2 = 4E\left\{E\left[X_n^2 \left(\sum_{k=1}^{n-1} a_{nk}X_k\right)^2 | \mathfrak{S}_{n-1}\right]\right\} \\
& = 4E\left\{\left(\sum_{k=1}^{n-1} a_{nk}X_k\right)^2 E[X_n^2 | \mathfrak{S}_{n-1}]\right\} \\
& = 4E\left[\left(\sum_{k=1}^{n-1} a_{nk}X_k\right)^2\right] \quad (\text{since } E[X_n^2 | \mathfrak{S}_{n-1}] = 1) \\
& = 4E\left[\sum_{k=1}^{n-1} a_{nk}^2 X_k^2 + 2 \sum_{j < k=1}^{n-1} a_{nj}a_{nk}X_jX_k\right] \\
& = 4\left[\sum_{k=1}^{n-1} a_{nk}^2 E(X_k^2) + 2 \sum_{j < k=1}^{n-1} a_{nj}a_{nk}E(X_jX_k)\right] \\
& = 4\left[\sum_{k=1}^{n-1} a_{nk}^2 E\{E(X_k^2 | \mathfrak{S}_{k-1})\} + 2 \sum_{j < k=1}^{n-1} a_{nj}a_{nk}E\{E(X_jX_k | \mathfrak{S}_j)\}\right] \\
& = 4 \sum_{k=1}^{n-1} a_{nk}^2
\end{aligned}$$

(since  $E(X_k | \mathfrak{S}_j) = 0$ ,  $X_k$  being a martingale difference sequence) and

$$\begin{aligned}
E[L_n^2] & = E[a_{nn}^2(X_n^2 - 1)^2] = E[a_{nn}^2(X_n^4 - 2X_n^2 + 1)] \\
& = a_{nn}^2[E\{E(X_n^4 | \mathfrak{S}_{n-1})\} - 2E\{E(X_n^2 | \mathfrak{S}_{n-1})\} + 1] \\
& = a_{nn}^2[c - 2 + 1] = a_{nn}^2[c - 1].
\end{aligned}$$

And lastly

$$\begin{aligned}
 E[K_n L_n] &= E\left[2X_n \left(\sum_{k=1}^{n-1} a_{nk} X_k\right) a_{nn} (X_n^2 - 1)\right] \\
 &= 2a_{nn} E\left\{E\left[\left(\sum_{k=1}^{n-1} a_{nk} X_k\right) (X_n^3 - X_n) \mid \mathfrak{S}_{n-1}\right]\right\} \\
 &= 2a_{nn} E\left\{\left(\sum_{k=1}^{n-1} a_{nk} X_k\right) E[(X_n^3 - X_n) \mid \mathfrak{S}_{n-1}]\right\} = 0
 \end{aligned}$$

(since  $E[X_n^3 \mid \mathfrak{S}_{n-1}] = E[X_n \mid \mathfrak{S}_{n-1}] = 0$ ). So

$$E[Y_n^2] = E[(K_n + L_n)^2] = E[K_n^2] + E[L_n^2] + 2E[K_n L_n] = 4 \sum_{k=1}^{n-1} a_{nk}^2 + a_{nn}^2 [c - 1].$$

We know that  $\sum_{k=1}^n Y_k = T_n$  converges in quadratic mean if and only if  $\sum_{n=1}^{\infty} E[Y_n^2] < \infty$  as  $\{Y_n\}$  is orthogonal.

But

$$\begin{aligned}
 \sum_{n=1}^{\infty} E[Y_n^2] &= \sum_{n=1}^{\infty} \left[4 \sum_{k=1}^{n-1} a_{nk}^2 + a_{nn}^2 [c - 1]\right] = 2 \sum_{n=1}^{\infty} \left[2 \sum_{k=1}^{n-1} a_{nk}^2 + a_{nn}^2\right] \\
 &\quad + [c - 3] \sum_{n=1}^{\infty} a_{nn}^2 = 2 \sum_{j,k=1}^{\infty} a_{jk}^2 + [c - 3] \sum_{n=1}^{\infty} a_{nn}^2 < \infty.
 \end{aligned}$$

So  $T_n$  converges in quadratic mean.

We have  $S_n = T_n + \sum_{k=1}^n a_{kk}$ .

Hence  $S_n$  converges in quadratic mean since  $\sum_{k=1}^{\infty} a_{kk} < \infty$ . To prove the almost sure convergence we have

$$\begin{aligned}
 T_n &= S_n - E[S_n] = \sum_{j,k=1}^n a_{jk} X_j X_k - \sum_{k=1}^n a_{kk} \\
 &= \sum_{j,k=1}^{n-1} a_{jk} X_j X_k - \sum_{k=1}^{n-1} a_{kk} + 2 \sum_{k=1}^{n-1} a_{nk} X_k X_n + a_{nn} (X_n^2 - 1)
 \end{aligned}$$

So

$$\begin{aligned}
 E[T_n \mid \mathfrak{S}_{n-1}] &= E\left[\left(\sum_{j,k=1}^{n-1} a_{jk} X_j X_k\right) \mid \mathfrak{S}_{n-1}\right] \\
 &\quad - \sum_{k=1}^{n-1} a_{kk} + 2E\left[\left(\sum_{k=1}^{n-1} a_{nk} X_k X_n\right) \mid \mathfrak{S}_{n-1}\right] + a_{nn} E[(X_n^2 - 1) \mid \mathfrak{S}_{n-1}] \\
 &= \sum_{j,k=1}^{n-1} a_{jk} X_j X_k - \sum_{k=1}^{n-1} a_{kk} + 2 \sum_{k=1}^{n-1} a_{nk} X_k E[X_n \mid \mathfrak{S}_{n-1}] + a_{nn} [E(X_n^2 \mid \mathfrak{S}_{n-1}) - 1] \\
 &= \sum_{j,k=1}^{n-1} a_{jk} X_j X_k - \sum_{k=1}^{n-1} a_{kk} = T_{n-1}.
 \end{aligned}$$

Thus it implies that  $\{T_n\}$  is a martingale. Now

$$\begin{aligned} E[|T_n|] &\leq \{E[T_n^2]\}^{1/2} = \left\{ \sum_{k=1}^n E(Y_k^2) \right\}^{1/2} \quad (\text{since } E(Y_j Y_k) = 0 \text{ for } j \neq k) \\ &\leq \left\{ \sum_{k=1}^{\infty} E(Y_k^2) \right\}^{1/2} = \left\{ 2 \sum_{j,k=1}^{\infty} a_{jk}^2 + (c-3) \sum_{k=1}^{\infty} a_{kk}^2 \right\}^{1/2} < \infty. \end{aligned}$$

Hence  $T_n$  converges almost surely and so  $S_n$  converges almost surely since  $S_n = T_n + \sum_{k=1}^n a_{kk}$  and  $\sum_{k=1}^{\infty} a_{kk} < \infty$ .

*Proof of Theorem 2.2.* We have

$$\begin{aligned} S_n &= \sum_{j,k=1}^n a_{jk} X_j X_k \\ &= \sum_{j=1}^n X_j \sum_{k=1}^{j-1} a_{jk} X_k + \sum_{k=1}^n X_k \sum_{j=1}^{k-1} a_{jk} X_j + \sum_{k=1}^n a_{kk} X_k^2 = K_n + L_n + M_n \quad (\text{say}) \end{aligned}$$

Now

$$\begin{aligned} E[K_n | \mathfrak{S}_{n-1}] &= \sum_{j=1}^{n-1} X_j \sum_{k=1}^{j-1} a_{jk} X_k + E \left[ X_n \sum_{k=1}^{n-1} a_{jk} X_k | \mathfrak{S}_{n-1} \right] \\ &= K_{n-1} + \left( \sum_{k=1}^{n-1} a_{nk} X_k \right) E[X_n | \mathfrak{S}_{n-1}] = K_{n-1}. \end{aligned}$$

So  $\{K_n\}$  is a martingale. Again

$$\begin{aligned} E[K_n^2] &= E \left[ \sum_{j=1}^n X_j^2 \left( \sum_{k=1}^{j-1} a_{jk} X_k \right)^2 \right] \\ &\quad + E \left[ \sum_{j \neq k} \sum_{l=1}^{j-1} \sum_{m=1}^{k-1} a_{jl} a_{km} X_j X_k X_l X_m \right] \\ &= E \left\{ E \left[ \sum_{j=1}^n X_j^2 \left( \sum_{k=1}^{j-1} a_{jk} X_k \right)^2 | \mathfrak{S}_{j-1} \right] \right\} \\ &\quad + \sum_{j \neq k} \sum_{l=1}^{j-1} \sum_{m=1}^{k-1} a_{jl} a_{km} E[X_j X_k X_l X_m] \\ &= E \left\{ \sum_{j=1}^n \left( \sum_{k=1}^{j-1} a_{jk} X_k \right)^2 \right\} \end{aligned}$$

(since for  $l < j, k; m < j, k \neq k$ ,  $E(X_j X_k X_l X_m) = 0$ , and  $E[X_j^2 | \mathfrak{S}_{j-1}] = 1$ )

$$= \sum_{j=1}^n E \left( \sum_{k=1}^{j-1} a_{jk}^2 X_k^2 + 2 \sum_{k < l=1}^{j-1} a_{jk} a_{jl} X_k X_l \right) = \sum_{j=1}^n \sum_{k=1}^{j-1} a_{jk}^2 < \infty.$$

Hence by martingale convergence theorem  $K_n$  converges almost surely. Similarly, it can be shown that  $L_n$  is also a martingale with respect to  $\mathfrak{F}_n$  and  $E[L_n^2] < \infty$ . So  $L_n$  also converges almost surely. Lastly

$$M_n = \sum_{k=1}^n a_{kk} X_k^2 = \sum_{k=1}^n a_{kk} (X_k^2 - 1) + \sum_{k=1}^n a_{kk}.$$

Here  $\sum_{k=1}^n a_{kk} (X_k^2 - 1)$  is a martingale and

$$E \left[ \left| \sum_{k=1}^n a_{kk} (X_k^2 - 1) \right| \right] \leq \sum_{k=1}^n |a_{kk}| E[|(X_k^2 - 1)|] \leq 2 \sum_{k=1}^n |a_{kk}| < \infty.$$

Thus  $\sum_{k=1}^n a_{kk} (X_k^2 - 1)$  converges almost surely.

Again

$$\sum_{k=1}^n a_{kk} \leq \left| \sum_{k=1}^n a_{kk} \right| \leq \sum_{k=1}^n |a_{kk}| \leq \sum_{k=1}^{\infty} |a_{kk}| < \infty.$$

So  $M_n$  converges almost surely. Hence the result follows.

*Proof of Theorem 2.3*

(a) Let  $S_n = \sum_{k=1}^n V_k X_k + \sum_{k=1}^n a_{kk} X_k^2 = P_n + M_n$  (say)

Now

$$E[P_n | \mathfrak{F}_{n-1}] = \sum_{k=1}^{n-1} V_k X_k + E[V_n X_n | \mathfrak{F}_{n-1}] = P_{n-1} + V_n E[X_n | \mathfrak{F}_{n-1}] = P_{n-1},$$

(since  $X_n$  is a martingale difference sequence).

So  $P_n$  is a martingale. Again

$$\begin{aligned} E[P_n^2] &= E \left[ \sum_{k=1}^n \left\{ \sum_{j=1}^{k-1} (a_{kj} + a_{jk}) X_j \right\}^2 X_k^2 \right] \\ &\quad + E \left[ \sum_{k \neq i}^n \sum_{j=1}^{k-1} \sum_{l=1}^{i-1} (a_{kj} + a_{jk})(a_{li} + a_{il}) X_i X_j X_k X_l \right] \\ &= \sum_{k=1}^n E \left[ E \left( \left\{ \sum_{j=1}^{k-1} (a_{kj} + a_{jk}) X_j \right\}^2 X_k^2 \mid \mathfrak{F}_{k-1} \right) \right] \\ &\quad + \sum_{k \neq i}^n \sum_{j=1}^{k-1} \sum_{l=1}^{i-1} (a_{kj} + a_{jk})(a_{li} + a_{il}) E[X_i X_j X_k X_l] \\ &= \sum_{k=1}^n E \left[ \left\{ \sum_{j=1}^{k-1} (a_{kj} + a_{jk}) X_j \right\}^2 \right] \end{aligned}$$

$$\begin{aligned}
&= \sum_{k=1}^n E \left[ \sum_{j=1}^{k-1} (a_{kj} + a_{jk})^2 X_j^2 + 2 \sum_{j < l=1}^{k-1} (a_{kj} + a_{jk})(a_{kl} + a_{lk}) X_j X_l \right] \\
&= \sum_{k=1}^n \sum_{j=1}^{k-1} (a_{kj} + a_{jk})^2 < \infty,
\end{aligned}$$

(using the fact that  $E[X_i X_j X_k X_l] = 0$  as proved in Theorem 2.2 and  $E[X_k^2 | \mathfrak{F}_{k-1}] = 1$ ).

So  $P_n$  converges almost surely using the martingale convergence theorem. Again

$$M_n = \sum_{k=1}^n a_{kk} X_k^2 = \sum_{k=1}^n a_{kk} (X_k^2 - 1) + \sum_{k=1}^n a_{kk}.$$

Here  $\sum_{k=1}^n a_{kk} (X_k^2 - 1)$  is a martingale and  $E \left[ \left| \sum_{k=1}^n a_{kk} (X_k^2 - 1) \right| \right] < \infty$  as shown in Theorem 2.2. Thus  $\sum_{k=1}^n a_{kk} (X_k^2 - 1)$  converges almost surely by the martingale convergence theorem.

Again

$$\sum_{k=1}^n a_{kk} \leq \left| \sum_{k=1}^n a_{kk} \right| \leq \sum_{k=1}^n |a_{kk}| \leq \sum_{k=1}^{\infty} |a_{kk}| < \infty.$$

So  $M_n$  converges almost surely and hence  $S_n$  converges almost surely.

$$\begin{aligned}
\text{(b)} \quad \sum_{k=1}^n V_k^2 &= \sum_{k=1}^n \left[ \sum_{j=1}^{k-1} (a_{kj} + a_{jk}) X_j \right]^2 \\
&= \sum_{k=1}^n \left[ \sum_{j=1}^{k-1} (a_{kj} + a_{jk})^2 X_j^2 + \sum_{j \neq l=1}^{k-1} (a_{kj} + a_{jk})(a_{kl} + a_{lk}) X_j X_l \right] \\
&= K_n + L_n \quad (\text{say})
\end{aligned}$$

Now

$$K_n = \sum_{k=1}^n \sum_{j=1}^{k-1} (a_{kj} + a_{jk})^2 (X_j^2 - 1) + \sum_{k=1}^n \sum_{j=1}^{k-1} (a_{kj} + a_{jk})^2.$$

Proceeding as in the proof of (a),  $\sum_{k=1}^n \sum_{j=1}^{k-1} (a_{kj} + a_{jk})^2 (X_j^2 - 1)$  is a martingale with  $E \left[ \left| \sum_{k=1}^n \sum_{j=1}^{k-1} (a_{kj} + a_{jk})^2 (X_j^2 - 1) \right| \right] < \infty$ . So  $\sum_{k=1}^n \sum_{j=1}^{k-1} (a_{kj} + a_{jk})^2 (X_j^2 - 1)$  converges almost surely.

Again

$$\sum_{k=1}^n \sum_{j=1}^{k-1} (a_{kj} + a_{jk})^2 < \infty \quad \text{as } n \rightarrow \infty.$$

So  $K_n$  converges almost surely. Now



$$L_n = \sum_{k=1}^n \sum_{j \neq l=1}^{k-1} (a_{kj} + a_{jk})(a_{kl} + a_{lk})X_j X_l$$

$$\leq \left[ \sum_{k=1}^n \sum_{j=1}^{k-1} (a_{kj} + a_{jk})^2 X_j^2 \right]^{1/2} \left[ \sum_{k=1}^n \sum_{l=1}^{k-1} (a_{kl} + a_{lk})^2 X_l^2 \right]^{1/2} \quad \text{for } j \neq l.$$

Hence  $L_n$  converges almost surely and so  $\sum_{k=1}^n V_k^2$  converges almost surely.

(c)  $\sum_{k=1}^n V_k^2 X_k^2 = \sum_{k=1}^n V_k^2 (X_k^2 - 1) + \sum_{k=1}^n V_k^2 = Q_n + R_n$  (say).

Now

$$E[Q_n | \mathfrak{S}_{n-1}] = E \left[ \sum_{k=1}^n V_k^2 (X_k^2 - 1) | \mathfrak{S}_{n-1} \right]$$

$$= \sum_{k=1}^{n-1} V_k^2 (X_k^2 - 1) + E[V_n^2 (X_n^2 - 1) | \mathfrak{S}_{n-1}]$$

$$= Q_{n-1} + V_n^2 [E(X_n^2 | \mathfrak{S}_{n-1}) - 1] = Q_{n-1},$$

so  $Q_n$  is a martingale.

Now

$$E[|Q_n|] = E \left[ \left| \sum_{k=1}^n V_k^2 (X_k^2 - 1) \right| \right] \leq \sum_{k=1}^n E[|V_k^2| | (X_k^2 - 1)|]$$

$$\leq \sum_{k=1}^n E\{E[|V_k^2| (|X_k^2| + 1) | \mathfrak{S}_{k-1}]\} = 2 \sum_{k=1}^n E[V_k^2]$$

$$= 2 \sum_{k=1}^n E \left[ \sum_{j=1}^{k-1} (a_{kj} + a_{jk}) X_j \right]^2 = 2 \sum_{k=1}^n E \left[ \sum_{j=1}^{k-1} (a_{kj} + a_{jk})^2 X_j^2 \right.$$

$$\quad \left. + 2 \sum_{j < l=1}^{k-1} (a_{kj} + a_{jk})(a_{kl} + a_{lk}) X_j X_l \right]$$

$$= 2 \sum_{k=1}^n \sum_{j=1}^{k-1} (a_{kj} + a_{jk})^2 E\{E[X_j^2 | \mathfrak{S}_{j-1}]\}$$

$$+ 4 \sum_{k=1}^n \sum_{j < l=1}^{k-1} (a_{kj} + a_{jk})(a_{kl} + a_{lk}) E\{E[X_j X_l | \mathfrak{S}_j]\} < \infty.$$

So  $Q_n$  converges almost surely.

We have already shown that  $R_n = \sum_{k=1}^n V_k^2$  converges almost surely. So  $\sum_{k=1}^n V_k^2 X_k^2$  converges almost surely.

### References

1. W. F. Stout, *Almost Sure Convergence*, Academic Press, New York, 1974.
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